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UNIVERSITY OF NOTRE DAME, NOTRE DAME INDIANA

Semi-Annual Status Report

to the

NATIONAL AERONAUTICS AND SPACE ADMINISTRATION

on

NASA Grant NSG-3048

ALTERNATIVES FOR JET ENGINE CONTROL*

November 1, 1983 - April 30, 1984

*This work has been carried out under the direction of

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ABSTRACT

This report documents the technical progress of researches under National Aeronautics and Space Administration Grant NSG-3048, entitled "Alternatives for Jet Engine Control", during the six-month period from November 1, 1983 to April 30, 1984. NASA Technical Officer for the work was Dr. Bruce Lehtinen, at Lewis Research Center. Dr. Michael K. Sain was director of the investigation at the University of Notre Dame.

Since the last report, Mr. Daniel Bugajski has completed an M.S. Thesis on the topics discussed therein. This work includes extensive numerical testing, over and above the previous progress document. Results indicate that optimal inputs contribute significantly to the process of calculating tensor approximations for nonlinear systems, and that the resulting approximations may be order-reduced in a systematic way.

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CHAPTER I

INTRODUCTION

Inherent in the design of any control scheme is a model of the physical system or plant. The plant itself can be modelled in basically two ways. First, laws of nature can be applied to the system in question. Many of these models are time tested and shall considered an adequate representation. Examples are the laws governing the mechanics of rigid bodies, and the set of equations governing the response of electric circuits. Or, second, if the system is far too complex for the relative simplicity of familiar equations, or if the governing scientific laws are too complicated to implement, system identification via excitation of the plant and measurement of the response can be performed [1]. It is this modelling technique we address here.

In the case of a linear approximation, this identification can be carried out without much computational difficulty. Unfortunately, because of system nonlinearities, in many instances the linear approximation, while adequate, leaves much room for improvement. In these cases, a model containing parameter estimates of higher degree terms (squared terms, cubic terms, and so forth) is an answer to characterizing the system nonlinearities.

Methods of calculating nonlinear models through the use of tensor algebraic ideas have been studied [2], and results have been good for academic examples [2] as well as for models of NASA's QCSE (Ouiet, Clean, Shorthaul, Experimental) jet engine [3]. As we shall see, the tensor approach is invaluable in that nonlinear problems are solved using linear techniques. A challenge of this tensor method, as with all general nonlinear methods, is the problem of complexity. Obviously the size of the model will increase with the

addition of states or control inputs; however a very noticeable increase in model size accompanies any increase in the degree of approximation. Presently, typical problems using square or cubic series approximations can be handled quite capably. As the need or desire to expand to fourth degree models, and beyond, surfaces, computational constraints will need increasing attention. Upper limits on addressable storage must be considered; and most matrix software routines have a vague upper limit on dimensions beyond which calculation becomes unreliable. In lieu of computational limitations, we can call on simple intuition to cite some disadvantages in the use of large models. A first obvious observation maintains that a model containing redundant information or one retaining useless information offers no advantages over the same model with the extraneous data disregarded. Both of these faults, however, are likely to occur as model size increases. A second, less compelling reason is that large models are simply more cumbersome than smaller, compact models.

In light of the above motives, clearly a scheme to reduce full size models could be nothing but beneficial to the modelling problem. This work then opens a pathway to the identification of such reduced systems by making use of a simple idea involving the comparison of squared errors. Finally, it is worth mentioning that state reduction is not addressed here, but rather reduction via the omission of selected terms in the series approximation is considered.

A brief review of the contents of the remaining chapters is appropriate. Chapter II presents some informal mathematical background. Among the items discussed are (1) overall problem formulation, (2) input design, (3) the sin-

gular value decomposition. Chapter III briefly summarizes the software package and the logical flow of each phase of the modelling/simulation procedure. In Chapter IV the scheme for model reduction is presented and discussed. It will be shown how the reduction method fits well into the present identification technique. Sample reductions and verifications are presented in Chapter V as a few example problems are studied. Finally, Chapter VI draws some conclusions and offers some pertinent suggestions.

CHAPTER II

PROBLEM BACKGROUND

The intent of this chapter is to provide a brief presentation of some of the principles on which the work in the following chapters is founded. The purpose here is not to relay detailed theory, but the material presented will be sufficient enough to gain a proper understanding of the problem.

The model structure formation, by means of a symmetric tensor method, is discussed in the first section along with a few ideas concerning the actual calculation of the model. Following this, the next section summarizes an information theoretic approach to input design used in the parameter identification. The final section presents an overview of the singular value decomposition because of its importance in the calculation of the model.

2.1 MODEL STRUCTURE

In a most general form, a nonlinear ordinary differential equation in ${\bf x}$ with input ${\bf u}$ can be written

$$\dot{x} = f(x, u)$$

where x is a real n-vector and u is a real m-vector.

Now suppose a Taylor series expansion is performed strough some operating point (x,u), then [2]

$$\hat{x} = f(\hat{x}, \hat{u}) = f(\overline{x}, \overline{u}) + \frac{\partial f}{\partial \hat{x}} | (\hat{x} - \overline{x}) + \frac{\partial f}{\partial \hat{u}} | (\hat{u} - \overline{u}) + \frac{\partial f}{\partial \hat{u}} | (\hat{x}, \overline{u})$$

$$+\frac{1}{2}\sum_{i=1}^{n} e_{i} (\hat{x}-\overline{x})^{T} \frac{\partial^{2} f_{i}}{\partial \hat{x}^{2}} \begin{vmatrix} (\hat{x}-\overline{x}) \\ (\overline{x},\overline{u}) \end{vmatrix}$$

$$+\sum_{i=1}^{n} e_{i} (\hat{x}-\bar{x})^{T} \frac{\partial^{2} f_{i}}{\partial \hat{x} \partial u} \begin{vmatrix} (\hat{u}-\bar{u}) \\ (\bar{x},\bar{u}) \end{vmatrix}$$

$$+\frac{1}{2}\sum_{i=1}^{n} e_{i} (\hat{\mathbf{u}}-\overline{\mathbf{u}})^{T} \frac{\partial^{2} f_{i}}{\partial \hat{\mathbf{u}}^{2}} \begin{bmatrix} (\hat{\mathbf{u}}-\overline{\mathbf{u}}) \\ (\overline{\mathbf{x}},\overline{\mathbf{u}}) \end{bmatrix}$$

+ . . .

where
$$e_i = \begin{bmatrix} \cdot \\ \cdot \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ \cdot \\ \cdot \end{bmatrix}$$
 \leftarrow ith position.

If the operating point $(\overline{x},\overline{u})$ is taken as a reference, then

$$x = \hat{x} - \bar{x},$$

$$u = \hat{u} - \overline{u},$$

and if

$$\bar{x} = f(\bar{x}, \bar{u}),$$

then

$$\dot{x} = \hat{x} - f(x, \overline{u}).$$

Substituting these into the expansion expression, we obtain

$$\dot{x} = \frac{\partial f}{\partial x} \left| \begin{array}{c} x + \frac{\partial f}{\partial \hat{u}} \\ (\overline{x}, \overline{u}) \end{array} \right| \frac{u}{(\overline{x}, \overline{u})}$$

$$+\frac{1}{2}\sum_{i=1}^{n} e_{i} x^{T} \frac{\partial^{2} f_{i}}{\partial \hat{x}^{2}} \begin{vmatrix} x \\ (\overline{x}, \overline{u}) \end{vmatrix}$$

+
$$\frac{1}{2} \sum_{i=1}^{n} e_{i} u^{T} \frac{\partial^{2} f_{i}}{\partial u^{2}}$$
 u

+ . . .

It is at this point where tensor algebra becomes important. Specifically the bilinear symmetric tensor product, \vee , [4] is used to form the products of the higher degree terms. If $x \in X$ and $u \in U$, then

We use these to write the differential equation as follows

fying the differential equation again

 $\dot{x} = L_{10}x + L_{01}u + L_{20}(x \vee x) + L_{11}(x \vee u) + L_{02}(u \vee u) + \dots$ where the L_{ij} make up the model. Note that the first and second subscripts correspond to the number of times x and u respectively are used in the symmetric tensor products associated with a given model partition L_{ij} . Simpli-

$$\dot{x} = \sum_{i=0}^{\infty} \sum_{j=0}^{\infty} L_{ij} \quad (x \lor x \lor x...) \quad (u \lor u \lor u...) .$$
i times

j times

Finally

$$\dot{x} = L z$$

where

$$L = [L_{10} L_{01} L_{20} L_{11} L_{02} L_{30} ...],$$

and z is a stacked vector of tensor term products:

It is interesting to note that the vector z is already reduced in one sense. Because of the properties of symmetric tensor algebra, redundant cross products are eliminated. For example

Thus, when x and u have dimension greater than one, cross products of individual elements contain redundancies and are hence eliminated. Some example instances are

 $x_1x_2u_1 = x_1u_1x_2$

= x2x1u1

 $= x_2u_1x_1$

= ulxlx2

 $= u_1x_2x_1$.

The existing software package includes an efficient ordering algorithm [2,3] which insures that redundant products are not calculated. It is clear that a significant amount of reduction in size, and hence, calculation is inherent is the use of symmetric tensor algebra.

We can now address the question concerning the calculation of the model, L, given a "black box" system. In our case the black box consists of a QCSEE digital simulation routine or a system of mathematical equations that represent a physical system.

Recall the equation

$$\dot{x}(t) = L z(t)$$

where $\dot{x}(t)$ is a vector of state derivatives and z(t) is a vector of tensor terms. Because L is the unknown, obviously we must know (or estimate) the values of $\dot{x}(t)$ and z(t) to identify the parameters contained in L. To get a reliable determination of how \dot{x} and z change with time, the system is perturbed from an equilibrium state, (\bar{x},\bar{u}) , to a trajectory (\hat{x},\hat{u}) where [5]

$$x = \hat{x} - \overline{x}$$

$$u = \hat{u} - \overline{u}.$$

The values of x(t) and u(t) and estimates of $\dot{x}(t)$ are then sampled and stored. From x(t) and u(t), z(t) is formed. The problem now appears:

$$\begin{bmatrix} \dot{x}_1(t_1) & \dot{x}_1(t_2) & \dots & \dot{x}_1(t_h) \\ \dot{x}_2(t_1) & \dot{x}_2(t_2) & \dots & \dot{x}_2(t_h) \\ \vdots & \vdots & \ddots & \vdots \\ \dot{x}_n(t_1) & \dot{x}_n(t_2) & \dots & \dot{x}_n(t_h) \end{bmatrix} = L \begin{bmatrix} z_1(t_1) & z_1(t_2) & \dots & z_1(t_h) \\ z_2(t_1) & z_2(t_2) & \dots & z_2(t_h) \\ \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \ddots & \vdots \\ z_p(t_1) & z_p(t_2) & \dots & z_p(t_h) \end{bmatrix}$$

where h is the number of samples and p is the number of terms in the nonlinear approximation; p is dependent on the number of states, n, the number of inputs, m, and the degree of approximation. We shall write this as X = LZ.

Fortunately, a very good routine for solving a problem of this form exists in the SPEAKEASY library of the IBM 370/3033. Using a singular value decomposition, discussed later, of Z, a least squares problem is solved [6] and L is calculated. The model then is ready for verification by means of digital simulation and comparison with true system solutions.

2.2 INPUT DESIGN

From the preceding section, it is clear that the model L is dependent on the excitation, (x(0),u(t)). Improper choice of perturbation might translate to an insufficient excitation of the nonlinearities, instability, or singularity. A careful method for choosing inputs could decrease the likelihood of encountering these defects. With this, the chance of identifying a "good" model, one with strong tracking ability and an acceptable region of stability, is then increased. In addition to better models, the capability of an input selection routine eliminates the need for an exhaustive search of input parameters.

For the purpose of this work, the theoretical detail of an input optimization derivation is not required. It is the intent here merely to present the basic members in the cost function calculation. For a theoretical and mathematical treatment, the reader is directed to [7]. The basic idea for the optimization comes from [8] and leads to the introduction of Fisher's Information Matrix, M. It is desired to get, in some sense, the minimum of a measure of M⁻¹. The formulation of Fisher's Information Matrix has its roots in probability, and many probabilistic rules and simplifications are used to express each element of M as

$$M_{i,j} = \sum_{k=0}^{N} \left[\frac{\partial}{\partial \theta_i} y(k) \right]^T R^{-1} \left[\frac{\partial}{\partial \theta_j} y(k) \right],$$

where

$$i = 1, 2, ... np,$$
 $j = 1, 2, ... np,$

and

N is the number of observation points, y(k) is the model output sequence, θ is a vector of parameters to be identified, R is the covariance matrix of the measurement of y(k).

Now for our problem let

$$y = \sum_{i=1}^{n} z_i$$

or for the discrete case

$$y(k) = L z(k).$$

Recall that θ is a vector of parameters to be identified, that is, each element of U. The partial derivatives can then be found. If

then the partial derivative of y = Lz with respect to some element of L, say ℓ_{ij} , is an nxl matrix with zeros in all but the i-th position where the entry is simply the j-th entry of z, z_i . Now let

$$R^{-1} = \begin{bmatrix} r_{11} & r_{12} & \cdots & r_{1n} \\ r_{21} & r_{22} & \cdots & r_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ r_{n1} & r_{n2} & \cdots & r_{nn} \end{bmatrix};$$

upon the calculation of each M_{ij} and the building of M, the following observation is made [7]

$$M = R^{-1} \otimes \overline{M}$$

where $\overline{\mathbf{M}}$ is symmetric and its upper triangular portion is shown in Figure 2.1.

Now that we have determined that \overline{M} and hence M are calculable, their role in the cost function is examined. The objective function chosen is

min
$$tr(M^{-1})$$
.

We use a property of the Kronecker product [9] to obtain

$$M^{-1} = R \otimes \overline{M}^{-1}$$

Another property allows

$$tr(M^{-1}) = tr(R)tr(\overline{M}^{-1}).$$

But if \overline{M} is an invertible matrix,

$$tr(\overline{M}^{-1}) = \sum_{i=1}^{p} \frac{1}{\lambda_i}$$

where the $\lambda_{\underline{i}}$ are the eigenvalues of \overline{M}_{\bullet} . Finally

$$\operatorname{tr}(M^{-1}) = \operatorname{tr}(R)(\sum_{i=1}^{p} \frac{1}{\lambda_{i}}).$$

It is worthwhile to note that, following from the above discussion, there are essentially two ways to generate the objective value; the eigenvalues of \overline{M} or the inverse of M must be calculated. In the case of real, symmetric matrices, these are similar in software requirements. Because we only need the trace, and not the entire inverse, we choose the eigenvalue approach.

•	•	•	•	•			
∑x ⊗ (u ∨ u)T	T(u ∨ u) ⊗ u∑	Σ(x v x) ⊗ (u v u) ^T	∑(x ∨ u) ⊗ (u ∨ u) ^T	_(n ∨ u) ⊗ (u ∨ u)√	•	•	
$\mathbf{x} \otimes (\mathbf{x} \vee \mathbf{u})^{\mathrm{T}}$	$\sum_{\mathbf{u}} \otimes (\mathbf{x} \vee \mathbf{u})^{\mathbf{T}}$	$\{(x \lor x) \otimes (x \lor u)^T$	∑(x ∨ u) ⊗ (x ∨ u) ^T		•	•	
$x \otimes (x \lor x)^T$	∑u⊗ (x ∨ x)T	∑(x ∨ x) ⊗ (x ∨ x) ^T			•		•
∑x ⊗ u ^T	∑u ⊗ u ^T				•	•	
X & XT			11		•	•	•
			Σ				

Figure 2.1. Upper Triangular Portion of the Matrix \overline{M}

This distinction becomes more important with increasing model size, which grows factorially with n, m, and degree of approximation. Despite the fact that matrix inversion is less acceptable, the current software makes provisions for both methods, for purposes of generality.

2.3 SINGULAR VALUE DECOMPOSITION

The discussion now turns to a brief presentation of the singular value decomposition (SVD). In this research effort, the SVD serves the important purpose of calculation of the generalized inverse of a matrix when the model is to be determined. The reader is directed to [10,11,12] for more formal and theoretically based discussions concerning the SVD and its computational implications and interpretations.

We first present the defining theorem of the SVD. If A is nxm with rank ρ , then there exist orthonormal matrices U (nxn) and V (mxm) such that

$$A = USV^{T}$$

where

T indicates transposition

and

S (nxm) is a "diagonal" matrix with diagonal elements si,

$$s_1 > 0$$
 $i = 1, 2, ... \rho$

$$s_i = 0$$
 $i = (\rho+1) \dots \min(n, m)$.

The columns of U are the left singular vectors and are the orthonormal eigenvectors of AA*. The columns of V are the right singular vectors and are the orthonormal eigenvectors of A*A. As expected the diagonal elements of S are the singular values. The non-zero elements of $\{s_i\}$ are the positive square roots of the eigenvalues of A*A (or AA*). This may seem to suggest a simple

way to calculate the singular values. This is not true, however, when finite precision arithmetic is involved. Because of the usually inexact representations due to truncation or rounding of numbers in a digital computer, the calculation of the eigenvalues of A*A can often lead to incorrect results concerning matrix rank [11,12].

Fortunately, SVD algorithms do exist that are computationally sound and not hindered by the above fault. Those that are generally considered among the best are the versions developed by Argonne National Laboratories. These include the SVD subroutine used by the SIMEQUAT algorithm of the SPEAKEASY library, IBM system 370/3033. The SIMEQUAT routine solves for the minimal solution x to the equation Ax=b. It does this by decomposing A and then finding the generalized inverse, A^I, (or A⁻¹ if A is square) of A. Given the generalized inverse, the problem becomes trivial. If

 $A = USV^{T}$

then

AI = VSIUT

and

 $x = A^{I}b$

becomes

v = vsIuTb.

Despite the existence of other methods for calculating the generalized inverse, A^I, the SVD remains the best. This is so because in problems where matrix rank is involved, as it is for matrix inversion, the SVD is the most reliable method of rank determination [11,12].

Finally, the concept of condition number is introduced. Very simply the condition number, k, is the ratio of the largest singular value to the smal-

lest non-zero singular value. That is

$$k = \frac{\max s_i}{\min s_i}.$$

There is a lot of abstract theory involving the interpretation of the condition number, but for the purposes herein, we will regard the condition number as a measure of "nearness to singularity".

2.4 REMARKS

This chapter presents the major ideas involved in the present modelling scheme. A sound, qualitative understanding of section one is most important, since throughout the remainder of this thesis, references to these ideas will be made. In section two, a summary derivation of the input design is presented. Here we are concerned not with mathematical rigor but rather with the idea that an input can be designed at all. We shall see the importance of this design when the examples are discussed.

CHAPTER III

SOFTWARE OVERVIEW

The intent of this chapter is simply to present the logical flow of the software package. This includes both the modelling and the simulation segments. Most of the CATNAP [3] software still exists, although in a more flexible form. In addition, software has been added to provide for the input optimization as well as for the model reduction test as presented in Chapter IV. The interactive modelling and simulation packages make use of two computers. They are the IBM 370/3033 and the DEC PDP 11/44.

3.1 MODELLING

The modelling segment of the present software package ties together each of the three ideas found in Chapter II. The IBM 370/3033 computer is used here in light of its computational power and its extensive support software libraries. The routine SUPRVIZE is an IBM command language (CLIST) program which governs the entire modelling segment. It is in SUPRVIZE that necessary libraries are made accessible and control is passed from loader routine to identification routine.

The first major duty of SUPRVIZE is compilation and execution of the proper lander routine. The loader routine (usually written in extended precision FORTRAN) has the chore of set up for identification and/or optimization. It is first determined whether a reduced model is to be identified. If so, a set of column numbers corresponding to the column numbers of the full model is read from the data set REDUCE. The products that correspond to these column numbers are omitted from further calculation. Then if an optimization is desired, the proper arrays are initialized and the minimization routine is

called. The minimization routine is supplied by IMSL (International Mathematical and Statistical Libraries) available on the IBM 370/3033. The IMSL routine requires a user supplied subroutine which calculates the cost function. In our case the cost function is computed using the eigenvalues of Fisher's Information Matrix as discussed in Section 2.2. Following the determination of inputs (whether by optimization or by user choice), the loader routine excites the system and forms the tensor term and state derivative matrices via sampling. These two matrices are loaded into the data set TEMP-FILE for later use.

After termination of the loader execution, the next duty of SUPRVIZE is to invoke the high level language SPEAKEASY [6]. The routines written in SPEAKEASY have two tasks: model calculation, and reduction test, if desired. As we have mentioned in Section 2.1, the model is calculated using a least squares approximation and singular value decomposition in the SIMEQUAT function of the SPEAKEASY library. Following output of the model parameters, the model is stored in data set MODEL. Now the reduction test can be performed. After completion of the test, if model reduction/re-optimization is desired, a set of column numbers is written to the data set REDUCE. SPEAKEASY is then exited and control is passed back to SUPRVIZE.

The final tasks of SUPRVIZE are simple. If the model is to be kept, the user is prompted for a model name and storage area (one of two partioned data sets). The model then is stored in the desired partioned data set with the given name. If another model is to be identified, the procedure restarts. Otherwise, unnecessary data sets are deleted and modelling is complete. See Figure 3.1 for a flow diagram of the entire modelling scheme.

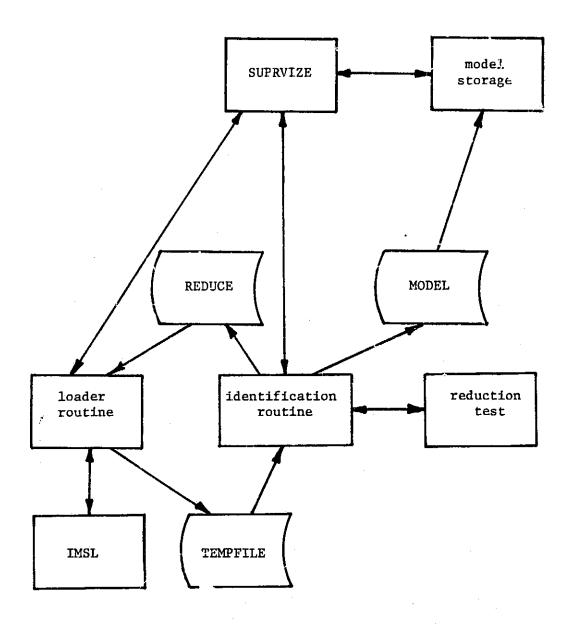


Figure 3.1. Logical Structure of Modelling Scheme

3.2 SIMULATION

Simulation of the nonlinear models can take place on either the IBM 370/3033 or the DEC PDP 11/44. For bulk tabular verification, the IBM 370/3033 is generally preferred due to its computational speed. Plot capability is handled with SAS (Statistical Aralysis Software). For plot comparison, though, it is preferred to use the DEC PDP 11/44. While computationally slower than the IBM, the DEC machine has high speed CRT plot capability with a hard copy unit for instant quality plots.

Simulation on the IBM 370/3033 is governed by the CLIST program COMPARE. COMPARE has several duties, the first of which is compilation of the simulation routine. The simulation routine (usually written in single precision FORTRAN) has the chore of solving the true system and integrating one or two models for a specified time interval. Furthermore, the simulation routine stores the data in the case that plotting is desired, and performs an error analysis between the models (if two models are simulated simultaneously). The error criteria is a simple ratic of mean square errors of the two models for each state.

After the simulation routine has been compiled, control passes back to COMPARE where the set up for simulation takes place. If a bulk tabular simulation is desired, COMPARE compiles and runs a short program that builds any specified table and stores it in data set SIMPNT. Upon exit from COMPARE, SIMPNT can be copied to a partitioned data set, TABLES, which acts as a library of simulation tables. If the desired simulation table is already built, COMPARE will ask for its name along with the names of the desired models. Following this, execution of the simulation routine is started. If the simulation is at a single point, the user is prompted for the input parameters.

Otherwise, the table is read for the input parameters of each simulation.

After all simulations have been completed, control is passed back to COMPARE.

If plotting is desired, control is passed to the SAS plotting routine. Otherwise the option to do another simulation is made. If simulation is complete,

COMPARE is exited. See Figure 3.2 for the logical flow of simulation on the

IBM 370/3033.

For simulation on the PDP 11/44, the models must first be transferred via magnetic tape from the IBM 370/3033. Once the model is on the DEC machine, an updated version of the CATNAP simulation routine is used. See Section 3.3 of [3] for an explanation of this simulation structure. Here we list only major changes to the old CATNAP software. They are

- comparison ability between any two models and the true solution (as opposed to strictly true solutions, linear model and some nonlinear model),
- instead of a separate Versatec hard copy routine, a hard copy unit which transfers the CRT image to paper is used,
- 3) the error criteria is now the ratio of mean square errors,
- 4) ability to blow up a portion of any plot for closer inspection of model trajectories.

In the writing of the current software, every effort was made to keep the programs general and flexible. See [15] for current IBM 370/3033 software listings. See also [16] for current listings of PDP 11/44 software.

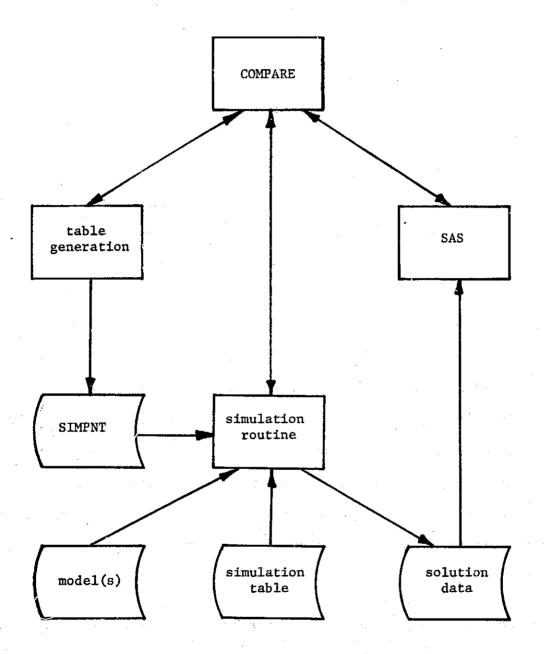


Figure 3.2. Logical Struction of Simulation on IBM 370/3033

CHAPTER IV

MODEL REDUCTION

In Chapter II some mathematical ideas were set forth that are at the center of the entire problem of nonlinear modelling via tensor parameterization. It is hoped in any nonlinear modelling exercise that the addition of higher degree terms will benefit the model in that the system nonlinearities will be more easily characterized by these higher degree terms or combinations there-But these models with the obvious advantage of higher degree lead us to a This challenge is the dimension or size of the model, p, and the challenge. arrays associated with its computation. Figure 4.1 shows how the size of model increases with the increase of the three determining factors, number of states, number of controls and degree. As the model gets larger, several issues may need to be examined, such as addressable storage limits, limits on computational reliability, and time of computation. Clearly a way of making the models easier to handle is very desirable. In short, if modelling is to proceed to higher degree terms, the models must have unnecessary monomials eliminated. It is to this end that we devote the remainder of this chapter to the introduction of one such size reduction method.

4.1 REDUCTION TECHNIQUE

As mentioned in past chapters, the model is calculated using a least squares approximation given times series state derivative data and time series tensor product data. Because this calculation uses a least squares approximation, and a truncation of the series approximation there will be an inherent error in the model. We can characterize the error, E, as

$$E = X - L Z$$
.

number of states	number of controls	degree of approximation	model size
2	2	1	2 x 4
2	2	2	2 x 14
2	2	3	2 x 34
2	2	4	2 x 69
2	3	1	2 x 5
2	3	2	2 x 20
2	3	3	2 x 55
2	3	4	2 x 125
2	4	1	2 x 6
2	4	2	2 x 27
2	4	3	2 x 83
2	4	4	2 x 209
3	2	1	3 x 5
3	2	2	3 x 20
3	2	3	3 x 55
3	2	4	3 x 125
3	3	1	3 x 6
3	3	2	3 x 27
3	3	3	3 x 83
<i>f</i>	3	4	3 x 209
3 3 3 3	4 4 4	1 2 3 4	3 x 7 3 x 35 3 x 119 3 x 315
4 4 4	2 2 2 2	1 2 3 4	4 x 6 4 x 27 4 x 83 4 x 209
4	3	1	4 x 7
4	3	2	4 x 35
4	3	3	4 x 119
4	3	4	4 x 315
4 4 4 4	4 4 4	1 2 3 4	4 x 8 4 x 44 4 x 164 4 x 474

Figure 4.1. Variation of Model Sizes with Various Parameters

Because \dot{X} and Z are time series data, E will become a time series matrix of state derivative errors. This simple calculation will become the base for the reduction method.

The idea of using an error matrix analysis is a modified approach from a treatment of nonlinear model reduction given by A.A. Desrochers [13].

Desrochers poses the problem with only a linear control input and in discrete time

$$x(k+1) = A F(k) + B u(k)$$

where

x is a state vector,

u is an input,

A and B constitute the model,

F is q-vector of states and state combinations, that is, F(k) =

To determine which columns 1 of A are most dominant, a modified state error vector $\tilde{\mathbf{e}}$ is formed

$$\tilde{e}_{j}(k) = x(k+1) - [A(,j)B] \begin{bmatrix} F(j,k) \\ ---- \\ u(k) \end{bmatrix} j = 1,2,...,q.$$

Note that the error is modified in the sense that it is formed using only one term at a time in addition to the control input. The state error vectors are then squared and summed for all h time points

$$\sum_{k=0}^{h-1} e_j(k) e_j(k).$$

Based on this squared error number, the most important terms can be kept and the others discarded.

Now in our more general problem, we have nonlinear input terms in addition to the linear terms, and we choose to test the effect of all terms in-

Henceforth, given a matrix A, we denote the ij-th element as A(i,j), the i-th column of A as A(i), and the i-th row as A(i,).

cluding those corresponding to the linear control inputs. (Recall Desrochers decided to keep all linear controls.) Our test error matrix E_1 is

$$E_{j} = \dot{x} - L_{j} Z_{j}, \quad j = 1, 2, ..., p$$

where

 E_j is the time series error matrix with the j-th term removed, p is the number of terms in the approximation, L_j is the model with the j-th column removed, Z_i is the time series tensor term data with the j-th row removed.

Note that now, in addition to the error incurred by the averaging of the least squares algorithm, there is error due the loss of the j-th term. We see that each column of E_j is the error vector at a certain time. Then we can use the same squared error calculation,

$$h-1$$
 T $\sum_{k=0}^{T} e_{j}(k) e_{j}(k)$ $j = 1,2,...,p$

where $E_j = [e_j(0) \mid e_j(1) \mid \dots \mid e_j(h-1)]$, to determine the effect of a certain term on the model. The influence of a term on the model is measured by checking the difference of the squared error of the full model and the squared error of the model shortened by that term. As an aside, it is interesting to note a philosophical difference between the problem formulation of Desrochers and this problem formulation. Desrochers chose to pick a significant term by including only that term in the error test. On the other hand, our reduction test checks a term's significance when that term is removed.

A simple example here will illustrate the calculations. Suppose we have a two state, one control system which we sample for four time points and calculate a linear approximation. Let

$$\dot{\mathbf{X}} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 1 \end{bmatrix}$$

and

$$Z = \begin{bmatrix} 0 & 1 & 2 & 3 \\ 1 & 2 & 3 & 0 \\ 2 & 3 & 0 & 1 \end{bmatrix}$$

and

$$L = \begin{bmatrix} 1 & 2 & 3 \\ 2 & -1 & 4 \end{bmatrix}.$$

First we calculate the state error matrix E of the full model, $E = \dot{X} - LZ$,

$$E = \begin{bmatrix} -7 & -12 & -5 & -5 \\ -5 & -9 & 3 & -9 \end{bmatrix}.$$

Now calculate the squared error for the full model, Y,

$$\Psi = \sum_{k=0}^{3} E^{T}(,k) E(,k) = [-7 \quad -5] \begin{bmatrix} -7 \\ -5 \end{bmatrix}$$

$$+ [-12 \quad -9] \begin{bmatrix} -12 \\ -9 \end{bmatrix}$$

$$+ [-5 \quad 3] \begin{bmatrix} -5 \\ 3 \end{bmatrix}$$

$$+ [-5 \quad -9] \begin{bmatrix} -5 \\ 9 \end{bmatrix}$$

$$= 439.$$

Following this calculation, we start checking the errors of the shortened models. We want first to calculate E_l which is the state error matrix with the first term of the model, x_l , and its corresponding data removed. According to the method, the first row of z is removed, the first column of L is removed, and the error calculations are repeated:

$$E_{1} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 1 & 3 & 1 \end{bmatrix}$$

$$-\begin{bmatrix} 2 & 3 \\ -1 & 4 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 & 0 \\ 2 & 3 & 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} -7 & -11 & -3 & 1 \\ -5 & -7 & 7 & -3 \end{bmatrix}.$$

$$\Psi_{1} = \sum_{k=0}^{3} E_{1}^{T}(,k) E_{1}(,k) = \begin{bmatrix} -7 & -5 \end{bmatrix} \begin{bmatrix} -7 \\ -5 \end{bmatrix}$$

$$= \begin{bmatrix} -11 & -7 \end{bmatrix} \begin{bmatrix} -11 \\ -7 \end{bmatrix}$$

$$+ \begin{bmatrix} -3 & 7 \end{bmatrix} \begin{bmatrix} -3 \\ 7 \end{bmatrix}$$

$$+ \begin{bmatrix} 1 & -3 \end{bmatrix} \begin{bmatrix} 1 \\ -3 \end{bmatrix}$$
$$= 312.$$

Then the relative error due to term x_1 is Ψ - Ψ_1 = 312-439 = -127.

Now we repeat the calculations for the removal of the second term in the model, x_2 . Here

$$E_{2} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 1 \end{bmatrix}$$

$$- \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix} \quad \begin{bmatrix} 0 & 1 & 2 & 3 \\ 2 & 3 & 0 & 1 \end{bmatrix}.$$

$$= \begin{bmatrix} -5 & -8 & 1 & -2 \\ -6 & -11 & 0 & -9 \end{bmatrix}.$$

The squared error number, Y2,

$$\Psi_2 = \sum_{k=0}^{3} E_2(k) E_2(k) = 332,$$

and the relative error is 332-439 = -107.

Finally the last term, corresponding to u, is checked. Thus

$$E_{3} = \begin{bmatrix} -1 & -3 & -5 & 1 \\ 3 & 3 & 3 & -5 \end{bmatrix},$$

$$\Psi_{3} = \sum_{k=0}^{3} E_{3}^{T}(,k) E_{3}(,k) = 88,$$

and the relative error is 88-439 = -351.

In most cases, examination and comparison of the computed relative errors will give a fair indication of which terms in the model are the most influential. Sometimes this is not the case and we must extend this squared error concept to look at more data. The extension is a simple and informative one.

Recall that each error matrix, E and the Ej's, is a time series matrix representation of the errors in the states. For example, in the previous

example the time series error in the first state of the full model is

$$E(1,) = [-7 -12 -5 -5],$$

the first row of E. Now use the squared error calculation for each individual state to get

$$\psi_1 = \sum_{k=0}^{h-1} E(i,k) E(i,k)$$
 $i = 1,2,...,n$.

This says that the error in the i-th state, ψ_{1} , is the sum of the squares of each of the entries of the i-th row of E. Furthermore note that the error in the entire system, Ψ_{1} is the sum of the ψ_{1} 's. That is

$$\Psi = \sum_{i=1}^{n} \psi_i,$$

and for the reduced cases

$$\Psi_{\ell} = \sum_{i=1}^{n} \psi_{\ell i},$$

where ℓ is the number of the term removed from the model. With these additional calculations, now, we can compare not only the entire system errors (the Ψ - Ψ_{ℓ}), but also the errors in the states ψ_{i} - $\psi_{\ell i}$.

To illustrate these calculations, we return to the example. We begin with the full model. The error in the first state is

$$\psi_1 = \sum_{k=0}^{3} E(1,k)E(1,k)$$

$$= (-7)^2 + (-12)^2 + (-5)^2 + (-5)^2$$

$$= 243_{5}$$

and the error in the second state is

$$\psi_2 = \sum_{k=0}^{3} E(2,k)E(2,k)$$

$$= (-5)^2 + (-9)^2 + (3)^2 + (-9)^2$$

= 196.

Thus

$$\Psi = \psi_1 + \psi_2 = 243 + 196 = 439.$$

After removing the first term in the model, the error in the first state is

$$\psi_{11} = \sum_{k=0}^{3} E_1(1,k)E_1(1,k)$$

$$= (-7)^2 + (-11)^2 + (-3)^2 + (1)^2$$

$$= 180,$$

and the error in the second state is

$$\psi_{12} = \sum_{k=0}^{3} E_1(2,k)E_1(2,k)$$

$$= (-5)^2 + (-7)^2 + (7)^2 + (-3)^2$$

$$= 132.$$

So

$$\Psi_1 = \psi_{11} + \psi_{12} = 180 + 132 = 312.$$

Now, as we calculated before, the change in the overall error is $\Psi_1 - \Psi =$ -127. In addition to this we check the change in the state errors. That is,

$$\psi_{11} - \psi_1 = 180 - 243 = -63$$

and

$$\psi_{12} - \psi_2 = 132 - 196 = -64.$$

Now, repeat the calculations when only the second term of the model is removed.

$$\psi_{21} = \sum_{k=0}^{3} E_2(1,k)E_2(1,k)$$

$$= (-5)^2 + (-8)^2 + (1)^2 + (-2)^2$$

$$= 94$$

$$\psi_{22} = \sum_{k=0}^{3} E_2(2,k)E_2(2,k)$$

$$= (-6)^2 + (-11)^2 + (0)^2 + (-9)^2$$

$$= 238.$$
Then
$$\Psi_2 = \psi_{21} + \psi_{22} = 94 + 238 = 332,$$

$$\Psi_2 - \Psi = 332 - 439 = -107,$$

and

$$\psi_{21} - \psi_1 = 94 - 243 = -149$$

 $\psi_{22} - \psi_2 = 238 - 196 = 42.$

Finally, the third term is removed and the calculations are performed. We get

$$\psi_{31} = 36$$
 $\psi_{32} = 52$

and $\Psi_3 = 88$, so $\Psi_3 - \Psi = -351$.

$$\psi_{31} - \psi_{1} = -207$$

$$\psi_{32} - \psi_2 = -144$$
,

which finishes the computation.

Arranging this data in a table will make it readable and usable. See Figure 4.2 for the reduction data of the sample problem. Naturally, as the model increases in size, the testing will increase to include all terms in the model approximation. These, then, are the simple calculations concerning the reduction method. When a reduction is done in the following chapter, an explanation of how to use these numbers will be presented.

Assume for now that the above test has indicated a set of terms are likely candidates for removal. There are several steps that can now be taken.

Full model system error: 439

Full model state errors: 243 196

Term: x₁

Change in system error: -127

Change in state errors: -63 -64

Term: x2

Change in system error: -107

Change in state errors: -149 42

Term: uj

Change in system error: -351

Change in state errors: -207 -144

Figure 4.2 Reduction Data for Sample Exercise

One option is to remove those terms from the model and retain the remainder of the terms thus producing the reduced model. This, although simple, is not a logical choice. For when we omit terms in a model, we are overlooking some dynamics. Regardless of their significance, these lost dynamics may be modelled adequately using the remaining terms, if they are given a chance. Hence, it is advantageous to regenerate the derivative and tensor data followed by a re-calculation of the now reduced model. But, by removing terms, we have changed the model structure and, intuitively, the system will need to be excited differently. Therefore, it would be a further advantage if a new input is designed.

The decision to re-optimize after terms are omitted seems sound, but it is by no means trivial. It was argued that a new input design is the sensible next step after the decision is made as to which terms are kept. But if the input changes (due to the optimization), the full model associated with this new excitation may contain influential terms at those places we have already made unavailable by reduction. Because of this difficulty, an assumption is made. It is assumed that a relatively insignificant term will remain relatively insignificant if the change in input excitation is not extreme. To ensure that the change in excitation is not too great, terms are not discarded in large groups. This is so, because it is hypothesized that a large change in model structure (that is, many terms removed) will cause the optimum input to move farther away from the present input parameters. Furthermore, the input optimization routine converges to a local minimum which will usually ensure the parameters do not move too much. To summarize, problems in reduction may be encountered if there is significant change between the input parameters of the model to be reduced and the input parameters of the reduced model.

As a result of the previous discussion, we have made a minor rule for reduction. It claims that terms to be omitted should not be discarded in large groups. There is another rule implicit in the preceding paragraph which is worth stating. The model to be reduced must have been identified at an input parameter set which is locally optimum. If it is not, optimization for the reduced model may change the input parameters significantly, and we have already argued that in general, this is not desired.

It is worthwhile to pause here to make some observations. First, by way of reasoning, a "multiple pass" reduction method has evolved. That is, an optimum input is chosen, the model is calculated, and the least significant terms in the model are determined. These terms are removed from further calculation, and an optimum input is re-chosen, the model is re-calculated, and so on. Obviously this process has a limit, and as the model becomes smaller, more care must be taken when choosing terms to be omitted. In some cases, the final pass is obvious. In these instances, the reduction test will indicate that all remaining terms are relatively significant. In other cases, the final pass is not so clear. Usually after a few passes, the model is adequately shortened, and the reduction test may indicate only a single term or two to be insignificant. It is in these cases that care must be taken when omitting terms. This is intuitively sensible since, as a model is made smaller, the dynamics in a single term are harder pressed to appear in the remaining terms. Because of this, model performance may suffer. It is in such cases that after a model has been reduced, some sample simulations should be run to check if model performance is still acceptable. If it is not, then the previous step was the final one. Otherwise reduction can continue. A second observation is that in requiring only a small change in input parameter design

from step to step in the reduction process, we have taken advantage of the fact that the IMSL optimization routine yields a local minimum. While it would be more advantageous to have a global minimization routine if full model identification was being done, in the case of iteratively reducing models, the local optimization is actually more desirable.

Returning to the reduction method, we can use the rules from the previous arguments to state the scheme in full:

- a) identify a model at an input which is locally optimum,
- b) determine a set of terms which seem to be the least significant.
- c) remove the terms in the set and re-optimize,
- d) identify reduced model,
- e) model simulation to test performance,
- f) if reduction data indicates more insignificant terms, go to step b), otherwise stop.

This simple method works quite well as the examples in the following chapter will show.

4.2 REMARKS

This chapter summarizes the present reduction method, including demonstration of the calculations and a step by step outline of the technique.

Difficulties arising from the combination of input design and term removal are discussed and a suitable compromise is obtained by the application of some basic rules. The following chapter contains concrete examples where the qualitative ideas of this chapter become clearer.

CHAPTER V

EXAMPLE REDUCTIONS

It is in this chapter that a few examples are studied to illustrate the power and usefulness of the model reduction method presented in the previous chapter. The first two systems were initially investigated in [2]. The first system is a two state, two control input set of nonlinear differential equations. The second system is a three state, three control set of strictly polynomic nonlinear differential equations. The final system is a three state, two input example. The first system will be modelled and reduced for second and third degree approximations. The other two examples will demonstrate second degree reductions.

The true test of a model's validity is borne out in simulation. Therefore much of the chapter is comprised of simulation data tables and plots.

With this in mind, we establish two general criteria for judging a reduced model. They are:

- a) the reduced model should approximate the full model of like degree of approximation,
- b) the reduced model should outperform the full model of less degree. Both of these are sensible goals and the examples herein do well with respect to them. Finally, it is worth restating a weak assumption necessary for model reduction. This is that the model should be an acceptable representation of the system. This is fairly obvious: since the motivation of a reduced model is to approximate the full model, we would hope the full model accurately describes the true system.

5.1 TWO STATE SYSTEM

The first system is given by the following equations

$$x_1 = u_2 \cosh(x_1 x_2) - e \quad \sinh(2x_1) - 3 \sinh(x_2),$$

$$u_1 u_2 \quad u_1$$

$$x_2 = e \quad \sinh(x_1) - e \quad u_1 \cosh(x_1^2) + \sinh(x_2).$$

These equations were studied in depth to demonstrate the power of the input optimization in identifying models that improve in quality as the degree of approximation increased. This study appears in Appendix A. It is from this report that "good" second and third degree models are taken as starting points for the reduction.

5.1.1 DEGREE TWO APPROXIMATION

For this system, the point of expansion is the origin because it is a stable equilibrium point. In the case of a second degree expansion, 14 model parameters corresponding to the tensor products

must be identified.

Now, to begin the reduction, a good full model that is locally optimum is necessary. The first appendix gives us such a model. That particular second degree model was identified using an initial state perturbation vector of

$$x_0 = (0.005, -0.005).$$

In addition the excitation for each input was taken to be a sum of two sinusoids:

$$u_1(t) = 0.025\sin(2\pi t \cdot \phi_1) + 0.025\sin(2\pi t \cdot 2\phi_1),$$

 $u_2(t) = 0.025\sin(2\pi t \cdot \phi_2) + 0.025\sin(2\pi t \cdot 2\phi_2).$

In this case $\phi = (\phi_1, \phi_2)$ was determined by optimization to be (0.7418, 0.9022) given the starting values (0.75, 1.0). The system was run for 4 seconds and 100 samples were taken at evenly spaced intervals.

Given this locally optimum model, the reduction test can be performed. See Figure 5.1.1 for the full model and its reduction test. The decision as to which terms should be omitted (or equivalently which terms should be kept) must now be made. Upon inspection of the reduction test, it should be noted that with the removal of any term, the change in the squared error number may increase or decrease significantly, or stay about the same. In this particular application, we choose to ignore the sign of the square error numbers and look upon their magnitudes as quantifiers of the effect of a term on the model. This is reasonable since removal of a term with a large change in squared error (either an increase or decrease) will most likely cause a large change in model performance. This, then, directs our attention to the terms with small changes in their errors. These terms are in some sense the least significant, so it is hoped that removal of them will have the least effect on model performance.

As a conservative first attempt, it is decided that error changes (both system and state errors) with magnitudes less than about 4 will be omitted.

The set of these terms is

x1x1, x1x2, x2x2, x1u2, x2u1, x2u2, u1u2.

Equivalently, the terms kept are

x1, x2, u1, u2, x1u1, u1u1.

Note that all linear terms were kept. In general, saving the linear terms is desired, since they hold the good local behavior.

Now, an input optimization is performed using only the six significant terms given by the reduction test
The system is again perturbed with initial conditions

```
ENTER INITIAL CONDITIONS FOR THE 2 STATES:

0.0050 -0.0050

ENTER SAMPLING PARAMETERS:

APPROXIMATION DEGREE

$ SAMPLE POINTS

INTERVAL BETWEEN SAMPLES

INTEGRATION STEPSIZE

2 100 0.0400 0.0050

SHOULD THE EXCITATION BE SINUSOIDAL OR COSINUSOIDAL?[S/C]:

IDENTIFICATION WILL BE ACKS WITH SINES.

ENTER THE NUMBER OF (CO)SINUSOIDS PER INPUT:

2 2

ENTER THE 4 INITIAL INPUT AMPLITUDES:
```

ENTER THE 4 INITIAL INPUT AMPLITUDES: 0.24999999999999996E-01 0.2499999999999996E-01

ENTER THE 2 FREQUENCY WEIGHTS: 0.7418 0.9022

SHOULD THE IDENTIFICATION BE PERFORMED USING DATA THAT IS REDUCED?[Y/N]:

ENTER OPTIMIZATION OPTION:

1 - NO OFTIMIZATIONS
2 - AMPLITUDES ONLY
3 - FREQUENCIES ONLY
4 - INITIAL CONDITIONS ONLY
5 - AMPLITUDES AND FREQUENCIES
6 - AMPLITUDES AND INITIAL CONDITIONS
7 - FREQUENCIES AND INITIAL CONDITIONS
8 - AMPLITUDES, FREQUENCIES, AND INITIAL CONDITIONS

DO YOU WISH TO NORMALIZE THE DATA?CY/NJ:

THE MATRIX OF SAMPLED MONOMIAL TERMS HAS 14 ROWS AND 100 COLUMNS.

NUMBER OF TIMES COST FUNCTION WAS EVALUATED: WOULD YOU LIKE TO SEE FLOTS OF THE INPUTS? IF SO, LINE UP THE CARRIAGE.

TSO SPEAKEASY III PI+ 7:16 PW FEBRUARY 22, 1984 :_SIZE=500;GET IDENT;IDENT;QUIT

EXECUTION STARTED

PARTITION NUMBER 1. PRTITION (A 2 BY 2 ARRAY)

Figure 5.1.1a Full Second Degree Model and Reduction Test (First Pass)

```
-2.0009 -2.9978
     ,9998 ,99982
 ...WITH EIGENVALUES:
VALUES (A VECTOR WITH 2 COMPONENTS)
-,50055+.86378i -,50055-,86378i
FARTITION NUMBER 2.
FRIITION (A 2 BY 2 ARRAY)
6.7287E-4 .99934
   -1.0003
                     -3.5733E-5
PARTITION NUMBER 3.
PRTITION (A 2 RY 3 ARRAY)
.0019401 -.080981 .034007
.0052751 9.3478E-4 -3.8931E-4
PARTITION NUMBER 4,
FRTITION (A 2 BY 4 ARRAY)
-4.0686 -.0062571 -.030538
-.018969 -.011543 -.01223
                                                               -,024848
                                                                 -.011688
PARTITION NUMBER 5.
PRITITION (A 2 BY 3 ARRAY)
-.033944 -6.0501E-4 .0049369
-.99912 .0066456 8.6412E-5
S (A 14 COMPONENT ARRAY)
9.7385 6.5841 6.4438 5.5233 3.802
1.5919 1.1872 .83463 .51123 .74939
THE MAXIMUM SINGULAR VALUE:
MAX = 9.7385
THE/MINIMUM NONZERO SINGULAR VALUE:
                                                                                         3.1743 2.3796
                                                                                                                          2,1075 1,8207
MIN = .51123
WANT TO TRY REDUCTION TEST?CY/NJ: Y
FULL SYSTEM ERROR: FULL STATE ERRORS:
                                              195.7138
171.4433
                                                                            24,2706
TERM: X1. COLUMN #: 1.
CHANGE IN SYSTEM ERROR: 187.8448
CHANGE IN STATE ERRORS: 145.1516
                                                                                       42+6933
                                      COLUMN #: 2.
RROR: 90.6012
PORS: 82.8124
TERM: X2. COLUI CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:
                                                                                         7,7888
TERM: U1. COLUMN #: CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:
                                                                                         8.0251
TERM: U2. COLUMN 4: 4. CHANGE IN SYSTEM ERROR: 39.0776 CHANGE IN STATE ERRORS: 39.0775
                                                                                         0.0001
TERM: X1.X1. COLUCHANGE IN SYSTEM ERROR:
```

Figure 5.1.1b Full Second Degree Model and Reduction Test (First Pass)

0.0306

COLUMN #:

```
CHANGE IN STATE ERRORS:
                                                     -0.6430
                                                                                 0.0736
TERM: X1.X2. COLUMN 1: 6.
CHANGE IN SYSTEM ERROR: -2.03
CHANGE IN STATE ERRORS: -2.01
                                                     -2,0255
                                                     -2.0145
                                                                               -0.0111
7.
-0.6560
                                                                               -0.0046
TERM: X1.U1. COLUMN $:
CHANGE IN SYSTEM ERROR: -1
CHANGE IN STATE ERRORS: -1
                                                 -122.4182
-122.3798
                                                                               -0.0384
TERM: X1.U2. COLUMN *:
CHANGE IN SYSTEM ERROR:
CHANGE IN STATE ERRORS:
                                                     -0.0639
0.0211
                                                                               -0.0849
TERM: X2.U1. COLUMN #:
CHANGE IN SYSTEM ERROR:
CHANGE IN STATE ERRORS:
                                                                                 0.0253
TERM: X2.U2. COLUMN #: 11.
CHANGE IN SYSTEM ERROR: 0.2700
CHANGE IN STATE ERRORS: 0.1913
                                                                                 0.0787
TERM: U1.U1. COLUMN #: CHANGE IN SYSTEM ERROR: -: CHANGE IN STATE ERRORS:
                                              N #: 12.
-15.4377
0.2414
                                                                             -15.6811
TERM: U1.U2. COLUMN #: 13.
CHANGE IN SYSTEM ERROR: -0.0333
CHANGE IN STATE ERRORS: -0.0192
                                                                               -0.0141
TERM: U2.U2. COLUMN #: 14.
CHANGE IN SYSTEM ERROR: -0.1777
CHANGE IN STATE ERRORS: -0.1797
                                                     -0.1797
                                                                                 0.0019
```

DO YOU WANT TO DISCARD ANY TERMS AND RE-OPTIMIZETCY/NJ: Y
HOW MANY TERMS WILL BE KEPT? 6

ENTER THE COLUMN NUMBERS OF COLUMNS WHICH ARE TO BE KEPT. WHEN ASKED, ENTER A AS AN ARRAY. A = ARRAY(6:1,2,3,4,8,12)

MANUAL MODE SPACE USED 63 K NOW, 72 K FEAK, SIZE 500 K DO YOU WISH TO SAVE THIS MODEL?[Y/N]:N DO YOU WISH TO IDENTIFY ANOTHER MODEL?[Y/N]:Y

Figure 5.1.1c Full Second Degree Model and Reduction Test (First Pass)

$$x_0 = (0.005, -0.005).$$

The control inputs are again given by

$$u_1(t) = 0.025\sin(2\pi t \cdot \phi_1) + 0.025\sin(2\pi t \cdot 2\phi_1),$$

$$u_2(t) = 0.025\sin(2\pi t \cdot \phi_2) + 0.025\sin(2\pi t \cdot 2\phi_2).$$

Now as starting frequencies for the optimization, we use the frequency set determined by the first pass optimimation, $\phi = (0.7418, 0.9022)$. The optimum frequencies are computed to be $\phi = (0.6499, 0.9014)$. Note that there is not much change in the optimum frequencies which is what we desire. The model is re-calculated and the reduction test again performed. See Figure 5.1.2 for the "second pass" output. Inspection of the reduction data indicates that the remaining terms are all fairly significant. So the model given in Figure 5.1.2 is the final reduced model. It contains 6 terms of a possible 14; a 57% reduction. Furthermore, by taking all the partial derivatives, the exact second degree approximation would be

$$L_{10} = \begin{bmatrix} -2 & -3 \\ 1 & 1 \end{bmatrix}, L_{01} = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix},$$

$$L_{20} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix},$$

$$L_{11} = \begin{bmatrix} -4 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix},$$

$$L_{02} = \begin{bmatrix} 0 & 0 & 0 \\ -1 & 0 & 0 \end{bmatrix}.$$

Obviously, the reduction method singled out the appropriate columns to be kept. In addition the error test for the individual states indicates the correct insignificant entries in each column. The reduction method seems to

```
ENTER INITIAL CONDITIONS FOR THE 2 STATES:
                0.0050
                                -0.0050
        ENTER SAMPLING PARAMETERS:
                       APPROXIMATION DEGREE
                       * SAMPLE POINTS
INTERVAL BETWEEN SAMPLES
INTEGRATION STEPSIZE
                 100
                            0.0400
                                           0.0050
        SHOULD THE EXCITATION BE SINUSOIDAL OR COSINUSOIDAL?ES/CJ:
        IDENTIFICATION WILL BE DONE WITH SINES.
        ENTER THE NUMBER OF (CO)SINUSCIDS PER INPUT:
         NTER THE 4 INITIAL INPUT AMPLITUDES: 0.24999999999999996E-01
       ENTER THE 2 FREQUENCY WEIGHTS: 0.7418 0.9022
        SHOULD THE IDENTIFICATION BE PERFORMED USING DATA THAT IS REDUCED?[Y/N]:
       ENTER OPTIMIZATION OPTION:

1 - NO OPTIMIZATIONS
2 - AMPLITUDES ONLY
3 - FREQUENCIES ONLY
                        - INITIAL CONDITIONS ONLY
- AMPLITUDES AND FREQUENCIES
- AMPLITUDES AND INITIAL CONDITIONS
- FREQUENCIES AND INITIAL CONDITIONS
- AMPLITUDES, FREQUENCIES, AND INITIAL CONDITIONS
       ENTER THE TRACE OF THE COVARIANCE MATRIX: 0.999999999999999954E-06
        ENTER THE OPTIMIZATION METHOD: 1,2,3 OR 4
       ENTER # OF SIGNIFICANT DIGITS FOR CONVERGENCE, MAXIMUM # OF FUNCTION CALLS, AND THE ZXMIN OPTION # (0,1,2,3)
2 200 2
       10 CALLS OF FCN....
       0.6872968D+00 0.9476237D+00
THE CONDITION NUMBER: 0.126D+05
OBJECTIVE FUNCTION VALUE: 0.1975D+00
         20 CALLS OF FCN....
       THE PARAMETERS:
0.6782293B+00 0.7468474D+00
THE CONDITION NUMBER: 0.118D+05
0BJECTIVE FUNCTION VALUE: 0.1918D+00
       30 CALLS OF FCN....
         0.6495584D+00
                                0.8921068D+00
       THE CONDITION NUMBER: 0.1070+05
OBJECTIVE FUNCTION VALUE: 0.17450+00
Figure 5.1.2a Reduced Second Degree Model and Reduction Test (Second Pass)
```

ORICANA DE

40 CALLS OF FCN....
THE PARAMETERS:
0.6478434D400 0.9013817D400
THE CONDITION NUMBER: 0.104D405
OBJECTIVE FUNCTION VALUE: 0.1719D400

50 CALLS OF FCN.....
THE PARAMETERS:
0,6498652D+00 0.9013841D+00
THE CONDITION NUMBER: 0.104D+05
0BJECTIVE FUNCTION VALUE: 0.1719D+00

60 CALLS OF FCN....
THE PARAMETERS:
0.6498653D+00 0.9013842D+00
THE CONDITION NUMBER: 0.104B+05
OBJECTIVE FUNCTION VALUE: 0.1719D+00

CONVERGENCE WAS ACHIEVED AND NO ERRORS OCCURRED.

DO YOU WISH TO NORMALIZE THE DATA?CY/NJ:

THE MATRIX OF SAMPLED MONOMIAL TERMS HAS 6 ROWS AND 100 COLUMNS.

NUMBER OF TIMES COST FUNCTION WAS EVALUATED: 63

THE OPTIMUM FREQUENCIES ARE: FREQ(1)= 0.650 FREQ(2)= 0.901 FREQ(3)= 0.000

WOULD YOU LIKE TO SEE PLOTS OF THE INPUTS? IF SO, LINE UP THE CARRIAGE.

TSO SPEAKEASY III PI+ 7:33 PM FEBRUARY 22, 1984 :_SIZE=500;GET IDENT;IDENT;QUIT

EXECUTION STARTED COLSKEPT (A 6 COMPONENT ARRAY) 1 2 3 4 8 12

PARTITION NUMBER 1. PRTITION (A 2 BY 2 ARRAY) -2.003 -3.0012 .9999 1.0001

...WITH EIGENVALUES:
VALUES (A VECTOR WITH 2 COMPONENTS)
-.50146+.86391 -.50146-.86391

FARTITION NUMBER 2. FRTITION (A 2 BY 2 ARRAY) 2.3743E-4 .9991 -1.0005 -1.5637E-5

Figure 5.1.2b Reduced Second Degree Model and Reduction Test (Second Pass)

```
PARTITION NUMBER 3.
PRITITION (A 2 BY 7 ARRAY)
0 0 0
0 0 0
```

PARTITION NUMBER 4.
FRIITION (A 2 BY 4 ARRAY)
-4.0049 0 0 0
.016344 0 0 0

PARTITION NUMBER 5.
PRITITION (A 2 BY 3 ARRAY)
-.016019 0 0
-.99954 0 0

S (A 6 COMPONENT ARRAY)
7.3342 2.012 2.2375 4.5216 5.5657 6.1157
THE MAXIMUM SINGULAR VALUE:
MAX = 7.3342
THE MINIMUM NONZERO SINGULAR VALUE:
MIN = 2.012

•

...AND THEIR RATIO: RTIO = 3.6452

WANT TO TRY REDUCTION TEST? CY/NJ: Y

FULL SYSTEM ERROR: FULL STATE ERRORS: 200.9931 176.7734 24,2198 TERM: X1. COLUMN #: 1. CHANGE IN SYSTEM ERROR: 236.6909 CHANGE IN STATE ERRORS: 194.4798 42,2111 TERM: X2. COLUMN #: 2. CHANGE IN SYSTEM ERROR: 67.8517 CHANGE IN STATE ERRORS: 60.9716 6.8801 TERM: U1. COLUMN #: 3. CHANGE IN SYSTEM ERROR: 11.0693 CHANGE IN STATE ERRORS: 0.0157 11.0535 TERM: U2. COLUMN #: CHANGE IN SYSTEM ERROR: -CHANGE IN STATE ERRORS: --16.1350 -16,1348 -0.0002 TERM: X1.U1. COLUMN *: 8. CHANGE IN SYSTEM ERROR: -120.4635 CHANGE IN STATE ERRORS: -120.5174 0.0539 TERM: U1.U1. COLUMN \$: CHANGE IN SYSTEM ERROR: -: CHANGE IN STATE ERRORS: 12. -15.9942 0.2470

DO YOU WANT TO DISCARD ANY TERMS AND RE-OPTIMIZE?[Y/N]: N

MANUAL MODE
SPACE USED 43 K NOW, 47 K PEAK, SIZE 500 K
DO YOU WISH TO SAVE THIS MODEL?CY/NJ:N
DO YOU WISH TO IDENTIFY ANOTHER MODEL?CY/NJ:N

Figure 5.1.2c Reduced Second Degree Model and Reduction Test (Second Pass)

have done about as well as could be expected. The true proof of a model's validity though lies in its simulation.

For the purpose of model verification, simulation consists of the excitation of the true system and the model at points about the point of identification. The control input excitation in this case is a single cosinusoid. That is

$$u_1(t) = \alpha_1 \cos(2\pi t \cdot \phi_1)$$
 $i = 1, 2, ..., m$

Furthermore, an initial condition for each state is required. The true system and each model are run for a specified time and sampled at evenly spaced points throughout the interval. If two models are simulated simultaneously, a mean square error (with respect to the true solution) analysis is performed. The error criterion, then, is just the ratio of the mean square errors of the two models for each state. For the following simulations, the error ratio is

$$R_{1} = \frac{MSE(MODELl_{1})}{MSE(LINEAR_{1})}, i = 1,2,...,n$$

where in this case MODEL1 refers to the i-th state of the six term reduced model. Obviously then, we desire the error ratios to be less than one. But, the following tabular data will show for some cases the error ratio between two models becomes very large. In these simulations it is necessary to check the normalized error data and closely examine the graphical output. Most of the time the large numbers are due to the fact that one model's trajectory lies on top of the true solution trajectory. Despite the fact that the other model may also be very close to the true solution, the error ratio becomes large.

Recall that two goals were set for performance evaluation of reduced models. The first is to outperform models of lesser degree, and the second is

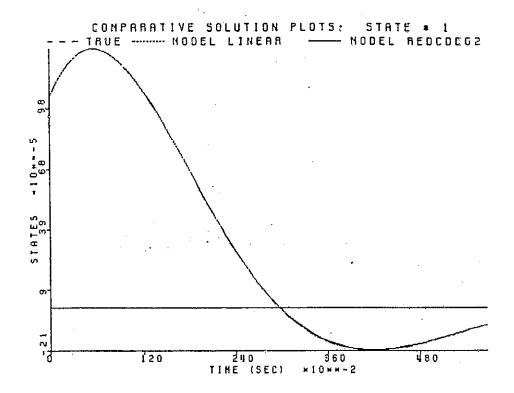
to approximate full model behavior. Tables 5.1.1 through Table 5.1.8 address the first criterion. In each of these tables, the heading MODEL1 refers to the reduced second degree model. The first three tables (Table 5.1.1 through Table 5.1.3) originally appear in [2] and test the model performance very close to the identification point. Note that throughout the trio of tables. according to the error ratios, the reduced model seems to be outperformed when the input amplitudes are zero. This behavior of the error criterion was explained above and a representative plot set is given in Figure 5.1.3.1 Representative plot sets of the simulation data tables follow each table. Note that even with the loss of terms, the second degree model exhibits exceptional behavior over the linear approximation. In Table 5.1.4, the model's region of validity is tested farther away from the origin as the input amplitudes are increased substantially over those in previous tables. Table 5.1.5 gives a set of simulations whose initial state conditions are (-0.01, -0.01), and whose input amplitudes and frequencies are randomly chosen in the ranges (-0.2, 0.2) and (2, 6) respectively. The following three tables (Tables 5.1.6 through Table 5.1.8) test the low frequency and d.c. behavior of the models. Table 5.1.6 has various small frequencies and steps. Note that it was previously determined in [2] that the true system is unstable in the first state when excited with steps of magnitude more than 0.1. Table 5.1.7 is another random table of smaller steps whose initial state conditions are in the range (-0.01, 0.01) and whose amplitudes are in the range (-0.025, 0.025). 5.1.8 has some larger steps with the initial state conditions randomly chosen in the range (-0.05, 0.05) and input amplitudes also randomly chosen in the range (-0.075, 0.075).

¹In all plots, if a curve is indistinguishable, it over-lays another curve.

Sŧ	INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
121345-6789012345678901234567890123456	0.001 0.001	0.050	00000000000000000000000000000000000000	0.748E+06 0.488E-05 0.549E-04 0.269E-05 0.452E-04 0.269E-05 0.729E-04 0.257E-03 0.162E-02 0.485E-03 0.257E-02 0.485E-03 0.3167E-02 0.485E-03 0.325E-03 0.374E-04 0.277E-05 0.194E-04 0.121E-05 0.194E-04 0.121E-03 0.325E-03 0.3371E-02 0.748E+06 0.325E-03 0.3371E-02 0.748E+06 0.325E-03 0.3371E-02 0.748E+06 0.325E-03 0.3371E-02 0.7581E-03 0.3371E-02 0.7581E-03 0.3371E-02 0.7581E-03 0.3371E-02 0.7581E-03 0.3371E-04 0.196E-05 0.3371E-02 0.7581E-03 0.3371E-04 0.196E-05 0.3371E-02 0.7581E-03 0.3371E-04 0.2581E-03 0.3371E-04 0.2759E-03 0.3371E-04 0.2759E-03 0.3871E-04 0.2759E-03 0.3871E-04 0.2759E-03 0.3871E-04 0.3871E-05

Table 5.1.1 Simulation Table for Linear Model versus Second Degree Reduced Model

No.



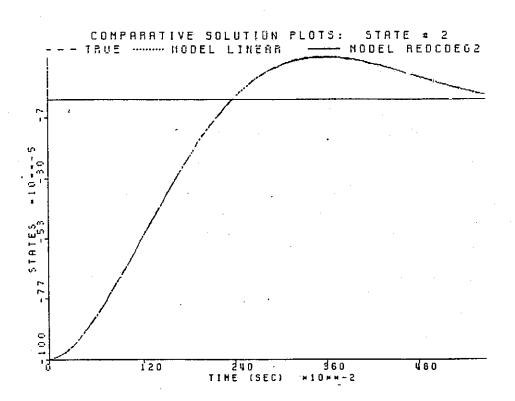
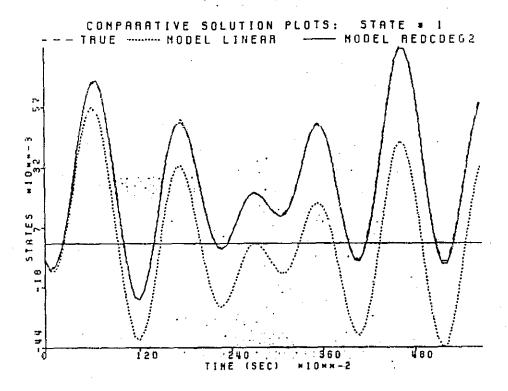


Figure 5.1.3 Simulation Number 10 of Table 5.1.1

OR POOR QUALITY.



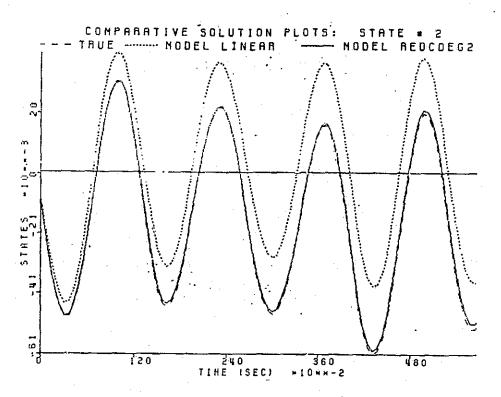
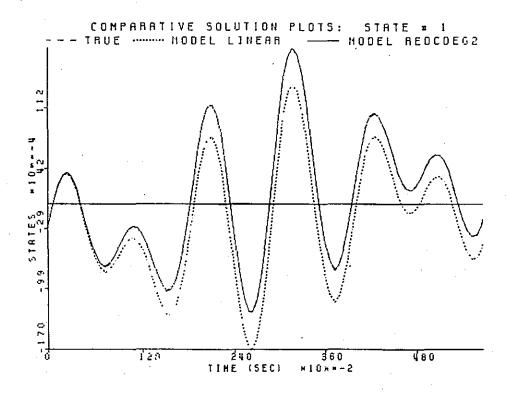


Figure 5.1.4 Simulation Number 25 of Table 5.1.1

3	S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
29	1 0.005 0.00	0.000 0.000 0.050 -0.050 -0.050 -0.050 -0.050 -0.150 -0.150 0.150 -0.150	0.7551.000 0.7551.000 0.7551.000 0.7551.000 0.7551.000 0.75551.000	805. 0.162E-04 0.170E-05 0.510E-05 0.811E-06 0.118E-05 0.811E-06 0.826E-04 0.159E-06 0.205E-03 0.203E-06 0.205E-03 0.203E-06 0.205E-03 0.203E-06 0.205E-03 0.203E-06 0.205E-03 0.203E-06 0.205E-03 0.203E-06 0.205E-03 0.203E-06 0.205E-03 0.203E-06 0.217E-05 0.374E-06 0.157E-05 0.374E-06 0.342E-05 0.327E-06 0.404E-03 0.217E-06 0.404E-03 0.217E-06 0.404E-03 0.219E-06 0.404E-03 0.331E-06 0.404E-05 0.331E-06 0.404E-05 0.331E-06 0.404E-05 0.331E-06 0.404E-05 0.331E-06 0.404E-05 0.331E-06 0.404E-05 0.553E-06 0.404E-05 0.703E-06 0.404E-05 0.703E-06 0.404E-06

Table 5.1.2 Simulation Table for Linear Model versus Second Degree Reduced Model



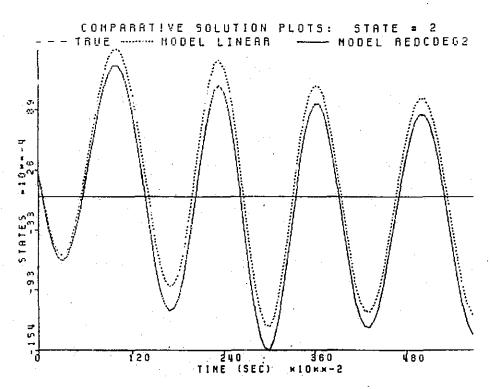
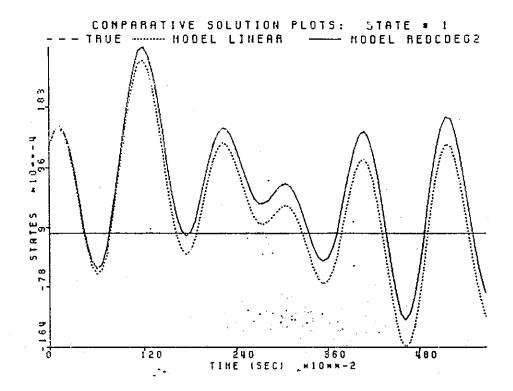


Figure 5.1.5 Simulation Number 29 of Table 5.1.2

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1 0.010 0.010 2 0.010 0.010 3 0.010 0.010 4 0.010 0.010 5 0.010 0.010 6 0.010 0.010 7 0.010 0.010 9 0.010 0.010 10 0.010 -0.010 11 0.010 -0.010 12 0.010 -0.010 13 0.010 -0.010 14 0.010 -0.010 15 0.010 -0.010 16 0.010 -0.010 17 0.010 -0.010 18 0.010 -0.010 19 -0.010 -0.010 20 -0.010 -0.010 21 -0.010 -0.010 22 -0.010 -0.010 23 -0.010 -0.010 24 -0.010 -0.010 25 -0.010 -0.010 27 -0.010 -0.010 28 -0.010 -0.010 29 -0.010 -0.010 30 -0.010 0.010 31 -0.010 0.010 32 -0.010 0.010 33 -0.010 0.010 34 -0.010 0.010	0.000 0.000 0.050 -0.050 -0.050 -0.050 -0.050 0.150 0.150 -0.150 0.150 -0.150 0.150 -0.150 -0.150 0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.150 0.150 0.150 -0.150 -0.150 0.050 -0.050 -0.050 -0.150 0.050 -0.150 0.050 -0.150 0.050 -0.150 0.050 -0.150 0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.150 -0.150 -0.150 -0.150 -0.150 -0.150 -0.150 -0.150 -0.150 -0.150 -0.150 -0.150 -0.150 -0.150	0.75 1.00 0.75 1.00	39.0 0.439E-04 0.166E-03 0.1364E-04 0.166E-03 0.134E-04 0.166E-03 0.134E-03 0.178E-02 0.150E-03 0.126E-02 0.570E-03 0.126E-02 0.570E-03 0.126E-02 0.570E-03 0.126E-02 0.570E-03 0.126E-02 0.570E-03 0.126E-02 0.570E-03 0.126E-02 0.534E-05 0.229E-04 0.180E-05 0.265E-04 0.180E-05 0.265E-04 0.184E-05 0.265E-04 0.184E-05 0.265E-04 0.184E-05 0.265E-04 0.115E-02 0.165E-02 0.165E-02 0.165E-02 0.165E-02 0.165E-02 0.165E-02 0.165E-02 0.165E-03 0.165E-02 0.165E-03
35 -0.010 0.010 36 -0.010 0.010	-0.150 -0.150 -0.150 0.150	0.75 1.00 0.75 1.00	0.365E-03 0.257E-02 0.883E-03 0.437E-02

Table 5.1.3 Simulation Table for Linear Model versus Second Degree Reduced Model



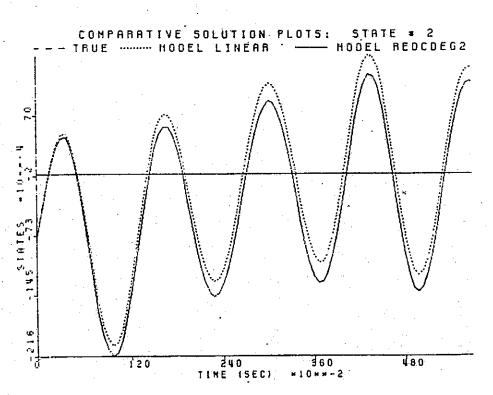
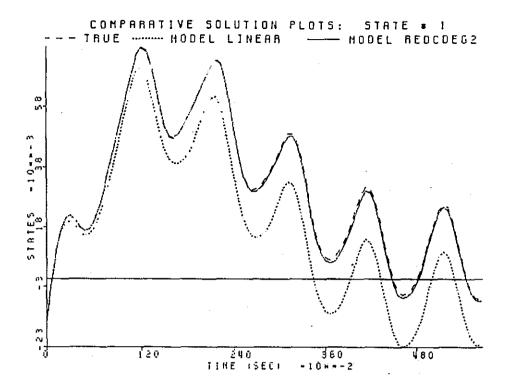


Figure 5.1.6 Simulation Number 14 of Table 5.1.3

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR EATIOS
2 0.025 0.025	0.100	2.00 1.00 2.00 1.00 2.00 1.00 2.00 1.00 2.00 1.00	0.566E-03 0.231E-02 0.467E-03 0.195E-02 0.467E-03 0.195E-02 0.4181E-03 0.115E-02 0.204E-03 0.115E-02 0.205E-03 0.115E-02 0.205E-03 0.115E-02 0.275E-03 0.862E-03 0.151E-03 0.850E-03 0.151E-03 0.850E-02 0.136E-03 0.359E-02 0.136E-03 0.442E-02 0.104E-02 0.514E-02 0.104E-02 0.514E-02 0.104E-02 0.514E-01 0.104E-02 0.101E-01 0.104E-02 0.101E-01 0.104E-02 0.101E-01 0.104E-02 0.101E-01 0.104E-02 0.101E-01 0.104E-02 0.101E-01 0.104E-02 0.101E-01 0.104E-02 0.101E-01 0.104E-02 0.101E-01 0.105E-01 0.101E-01 0.101E-01 0.589E-01 0.101E-01 0.589E-01 0.101E-01 0.510E-01 0.101E-01 0.510E-01 0.101E-01 0.510E-01 0.101E-01 0.510E-01 0.101E-01 0.510E-01 0.101E-01 0.510E-01 0.101E-01 0.510E-01 0.101E-01 0.510E-01

Table 5.1.4 Simulation Table for Linear Model versus Second Degree Reduced Model



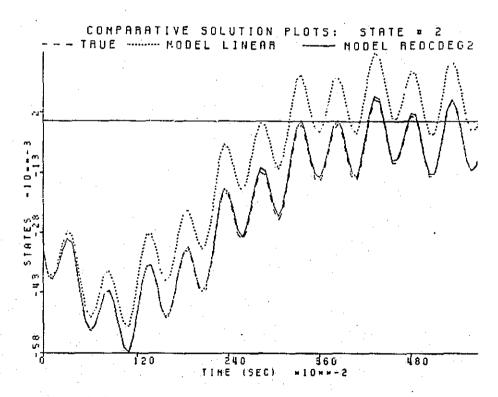
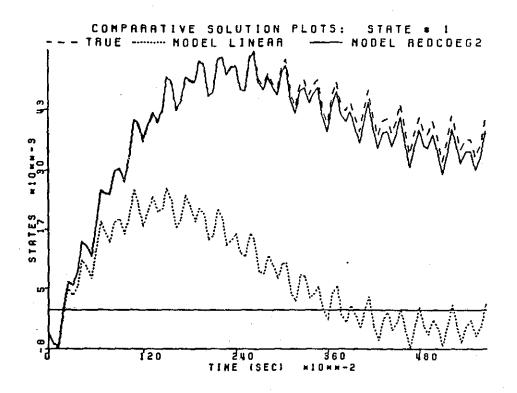


Figure 5.1.7 Simulation Number 13 of Table 5.1.4

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1	0.063	6977411108804519488408520977534765320260183431224268542455451740180352221585209775347653202601834312224268542454545174899925885965775496582333324058406174339082545234522454525345523553445555236235242424553454545452335534455552362352424242534545	0.552E-04 0.315E-04 0.273E-03 0.215E-03 0.362E-03 0.248E-03 0.362E-03 0.408E-04 0.113E-04 0.970E-04 0.977E-04 0.216E-01 0.721E-03 0.137E-02 0.458E-02 0.125E-02 0.107E-02 0.125E-02 0.107E-02 0.125E-02 0.107E-03 0.176E-02 0.125E-02 0.263E-02 0.129E-03 0.263E-02 0.129E-03 0.265E-02 0.142E-03 0.296E-03 0.335E-04 0.322E-04 0.105E-04 0.322E-04 0.105E-04 0.322E-04 0.105E-04 0.322E-04 0.105E-04 0.115E-02 0.117E-03 0.274E-03 0.127E-03 0.274E-03 0.127E-03 0.274E-03 0.127E-03 0.274E-03 0.127E-03 0.115E-02 0.662E-04 0.900E-03 0.127E-03 0.115E-02 0.623E-03 0.111E-02 0.373E-04 0.111E-02 0.250E-04 0.201E-03 0.147E-02 0.581E-02 0.355E-03 0.111E-02 0.359E-03 0.111E-02 0.312E-03 0.536E-02 0.496E-03 0.536E-02 0.496E-03 0.536E-02 0.496E-03 0.536E-02 0.496E-03 0.536E-02 0.474E-03 0.448E-03 0.474E-03 0.448E-03 0.474E-03 0.448E-03 0.474E-03 0.536E-02 0.359E-05 0.426E-03 0.359E-05 0.426E-03 0.672E-03 0.766E-03 0.672E-03 0.766E-03 0.672E-03 0.766E-03 0.672E-03 0.766E-03 0.672E-03 0.766E-03 0.672E-03 0.766E-03

Simulation Table for Linear Model versus Second Degree Table 5.1.5 Reduced Model



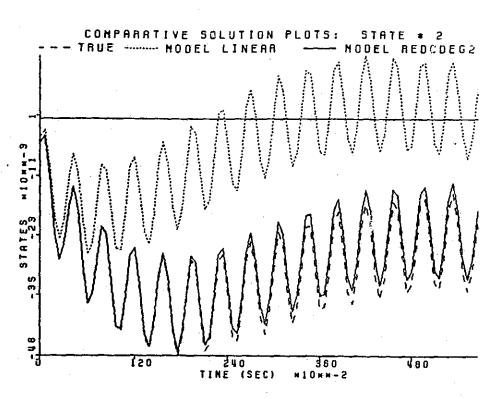
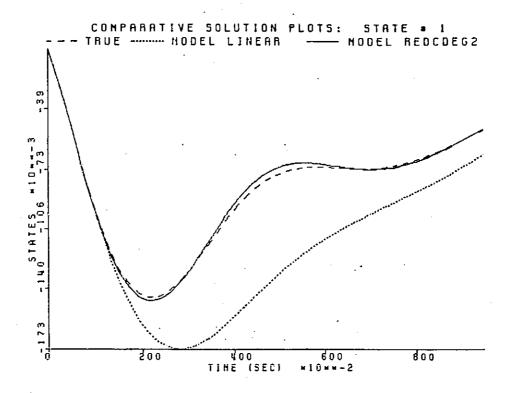


Figure 5.1.8 Simulation Number 37 of Table 5.1.5

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1	0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.075 -0.050 0.050 -0.050 0.050 -0.050 0.050 -0.075 0.010 -0.075 0.010 -0.075	0.001 0.002 0.001 0.002 0.001 0.002 0.001 0.002 0.001 0.002 0.001 0.002 0.001 0.002 0.001 0.002 0.001 0.002 0.001 0.002 0.001 0.002 0.001 0.002 0.003 0.	0.827E-04 0.769E-04 0.754E-04 0.595E-04 0.545E-04 0.597E-04 0.218E-34 0.180E-02 0.683E-02 0.548E-02 0.644E-02 0.548E-02 0.197E-02 0.165E-02 0.359E-01 0.487E-01 0.354E-01 0.487E-01 0.412E-01 0.216E-01 0.103E-03 0.115E-03 0.109E-03 0.115E-03 0.124E-04 0.747E-04 0.724E-04 0.747E-04 0.521E-04 0.507E-04 0.213E-02 0.563E-02 0.564E-01 0.469E-01 0.724E-04 0.507E-04 0.523E-04 0.177E-04 0.507E-03 0.114E-03 0.179E-03 0.115E-03 0.109E-03 0.117E-03

Table 5.1.6 Low Frequency Table for Linear Model versus Second Degree Reduced Model



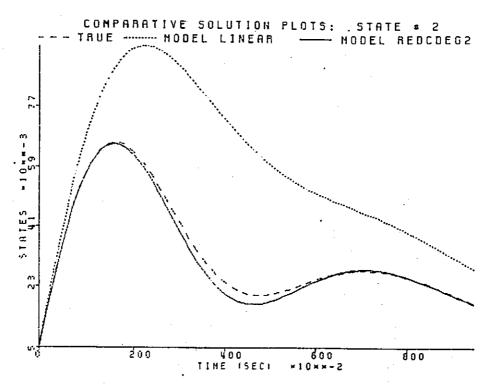


Figure 5.1.9 Simulation Number 15 of Table 5.1.6

PROBLEM SUMMARY CONFIGURATION: TRUE:LIMEAR.MODEL1 OF STATES: 2 OF INPUTS: 2 OF TERMS IN MODEL 1: 6 DEGREE OF APPROXIMATION: 2 SIMULATION WITH COSINE

水水水水水水水水水

FREQUENCIES **ERROR RATIOS AMPLITUDES** S# INITIAL CONDITIONS 0.506E-05 0.488E-05 0.144E-05 0.1122-05 0.207E-06 0.123E-06 0.00 0.010 -0.021 0.00 -0.003 0.010 0.001 -0.002 0.008 0.903 -0.006 -0.015 0.002 0.023 0.00 0.00 0.00 ♣ 0.00B 0.463E-04 0.463E-04 0.463E-05 0.249E-05 0.249E-05 0.249E-05 0.252E-05 0.441E-05 0.166E-05 0.294E-06 0.294E-05 0.294E-05 0.294E-05 0.294E-05 0.294E-05 0.294E-05 0.294E-05 0.153E-05 0.153E-05 0.153E-05 0.153E-05 0.153E-05 0.153E-05 0.153E-06 0.153E-06 0.153E-06 0.153E-07 0.167E-06 0.167E-06 0.173E-07 0.167E-06 0.173E-07 0.167E-06 0.173E-07 0.167E-07 -0.022 0.003 0.015 0.00 0.00 0.010 -0.003 0.00 Š -0.009 0.002 0.010 0.00 -0.006 0.012 0.008 0.002 -0.0130.00 0.00 67 0.005 0.005 0.00 0.00 -0.002 0.014 0.00 0.011 0.00 8 0.00 0.00 -0.005 -0.004 0.002 -0.008 -0.012 0.007 0.00 0.00 -0.01B 10 -0.000 11 12 13 0.006 0.00 0.00 -0.005 -0.003 0.00 0.00 -0.004 -0.0140.000 -0.007 0.00 -0.001 -0.008 -0.015 -0.002 0.003 0.00 0.00 0.008 0.009 -0.005 -0.004 -0.001 0.00 9.00 0.004 ĪŚ 16 17 -0.0010.004 0.00 0.00 -ŏ.ŏŏŝ 0.001 -0.007 0.00 0.00 -0.008 0.001 0.005 0.007 -0.008 0.00 Ĭ8 0.003 0.008 0.00 0.00 0.009 -0.004 -0.0240.00 19 20 21 22 0.00 0.009 0.002 0.016 0.00 0.00 -0.003 0.00 0.006 0.003 -0.009 0.008 0.021 0.023 0.00 0.00 -0.002 0.00 -0.003 -0.021 -0.007 0.00 -0.011 -0.023 0.014 0.008 -0.002 0.00 0.016 0.00 0.001 0.00 -0.005 0.000 -0.0240.00 -0.024 0.024 0.00 -0.004 0.014 0.00 ŏ.ŏŏ 0.006 -0.006 0.001 -0.011 0.016 -0.019 0.00 0.008 0.00 -0.003 0.014 0,009 0.010 0.00 0.00 0.022 -0.024 0.021 0.00 0.008 -0.025 0.00 0.003 0.009 0.00 -0.0100.004 --0.002 0.002 0.020 0.00 0.00 -0.009 0.004 -0,021 -0.009 0.00 0.00 -0.011 -0.017 0.00 35 -0.00B 0.004 -0.005 0.00 0.005 -0.008 0.003 0.00 0.00 36 0.008 0.007 0.00 0.00 -0.006 0.006 0.008 -0.017 -0.008 0.018 0.00 0.00 0.000 -0.016 0.019 0.020 0.018 0.012 0.00 0.00 0.00 0.00 0,006 40 0.017 0.00 0.00 -0,008 -0.004 41 0.014 0.022 -0.017 -0.012 0.00 42 43 44 -0.007 -0.009 0.00 -0.006 -0.008 0.006 -0.012 -0.007 -0.001 0.003 0.00 0.00 0.00 0.00 0.00 0.007 0.00 45 -0,005 0.005 0.003 -0.022 -0.004 0.002 -0.005 0.00 0.00 46 0.020 0.015 0.00 47 -0.004 -0.005 0.00 0.00 0.00 48 -0.0040.003 -0.009 0.00 -0.010 -0.001 49 0.011 -0.007

Table 5.1.7 Step Response Table for Linear Model versus Second Degree Reduced Model

0.001

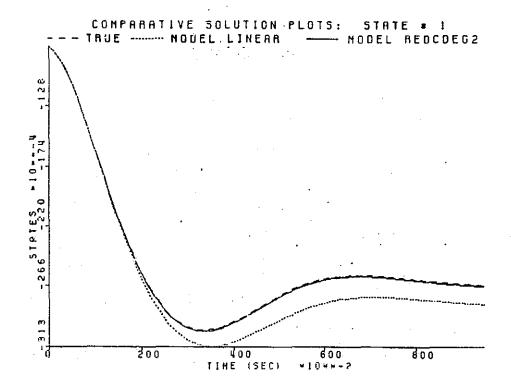
-0.006

50

0.007 -0.003

0.00

0.00



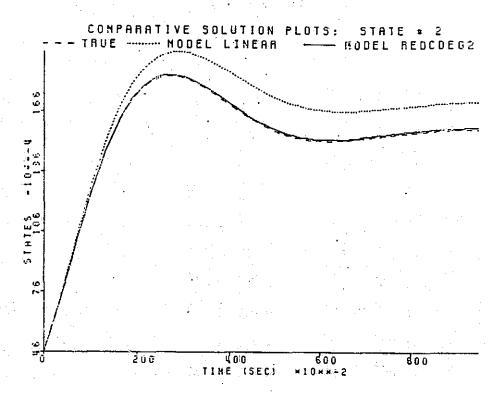
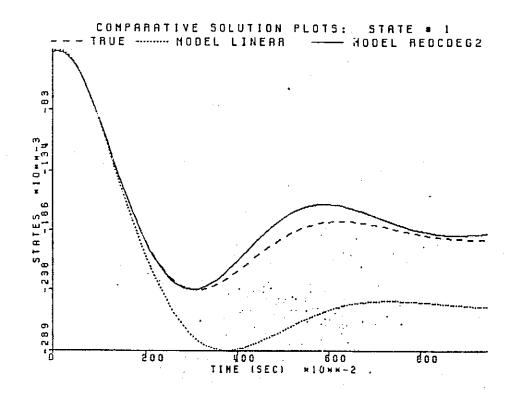


Figure 5.1.10 Simulation Number 35 of Table 5.1.7

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1	-0.062 -0.050 -0.063 -0.063 -0.005 -0.0642 -0.005 -0.022 -0.007 -0.024 -0.024 -0.0257 -0.024 -0.024 -0.024 -0.0256 -0.024 -0.029 -0.024 -0.029 -0.033 -0.029 -0.035 -0.014 -0.036 -0.057 -0.015 -0.013 -0.018 -0.022 -0.019 -0.022 -0.019 -0.022 -0.019 -0.022 -0.019 -0.021 -0.021 -0.021 -0.021 -0.023 -0.021 -0.023 -0.024 -0.023 -0.025 -0.023 -0.025 -0.023 -0.021 -0.023 -0.024 -0.023 -0.025 -0.023 -0.024 -0.023 -0.025 -0.023 -0.024 -0.023 -0.024 -0.023 -0.024 -0.024 -0.033 -0.0249 -0.025 -0.025 -0.026 -0.025 -0.027 -0.029 -0.029 -0.029 -0.029 -0.033 -0.029 -0.033		0.451E-03 0.520E-03 0.451E-05 0.344E-05 0.339E-02 0.426E-02 0.350E-05 0.344E-05 0.581E-03 0.570E-04 0.724E-03 0.743E-03 0.622E-02 0.733E-03 0.240E-03 0.245E-03 0.240E-03 0.245E-03 0.143 0.245E-03 0.143 0.245E-03 0.143 0.245E-03 0.143 0.245E-03 0.143 0.253E-04 0.190E-03 0.253E-06 0.190E-03 0.253E-06 0.190E-03 0.395E-03 0.143 0.395E-04 0.127E-04 0.253E-06 0.140 0.395E-03 0.106E-05 0.395E-03 0.1079E-03 0.529E-03 0.1079E-03 0.529E-03 0.1079E-04 0.554E-01 0.179E-06 0.347E-06 0.179E-06 0.347E-06 0.179E-01 0.885E-01 0.779FE-01 0.885E-01 0.179E-02 0.785E-03 0.121E-03 0.795E-03 0.1225E-01 0.561E-05 0.135E-01 0.561E-05 0.135E-01 0.136E-01 0.785E-02 0.785E-02 0.106E-01 0.136E-01 0.785E-02 0.785E-02 0.106E-01 0.136E-01 0.793E-01 0.561E-05 0.127E-02 0.785E-02
48 -0.036 0.018 49 0.026 0.017 50 -0.029 0.038	-0.074 0.031 0.061 0.006 -0.023 -0.055	0.00 0.00 0.00 0.00 0.00 0.00	0.474 0.514 0.920E-01 0.119 0.254E-05 0.593E-06
UV TU:VZ7 V:V30	A+659 A+659	7 5 00 V 100	A15946_A9 A19196_A0

Table 5.1.8 Step Response Table for Linear Model versus Second Degree Reduced Model



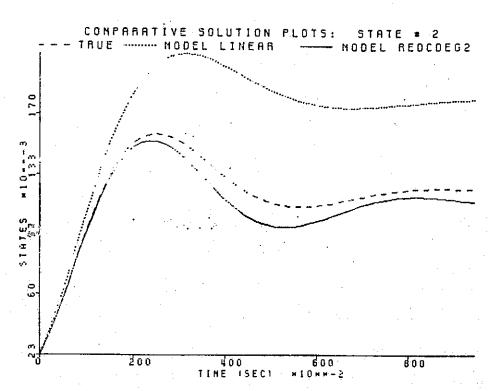


Figure 5.1.11 Simulation Number 48 of Table 5.1.8

From the tables and plots it is obvious that the second degree reduced model performs very well with respect to the linear model. Clearly, the loss of the eight terms in the model has not caused model performance to deteriorate significantly. In order to better check the reduced model's degradation of tracking ability, it is compared against the full second degree model. Tables 5.1.9 through 5.1.13 and the corresponding lots will show that good model behavior is preserved despite the loss of more than half the full model's terms. In each case now, we do not expect to outperform the full model, rather we just hope to approximate it. The error ratio, now

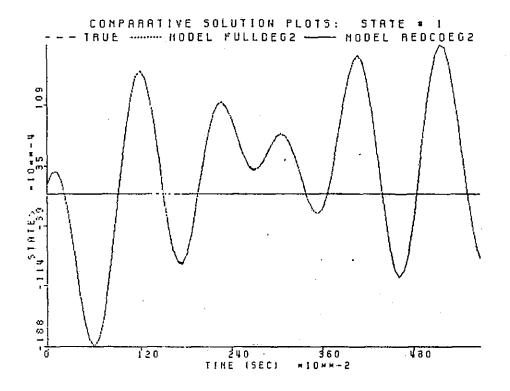
$$R_{i} = \frac{MSE(MODEL2_{i})}{MSE(MODEL1_{i})}$$
, $i = 1,2,...,n$,

(where MODEL1 and MODEL2 refer to the full and reduced models respectively) is expected to be around unity. This may or may not be true, again depending on the relative closeness of a model's solution to the true solution, so we rely on the graphical data. Table 5.1.9 is a table of simulations close to the origin. The error ratios meet their expectation and the plots in Figures 5.1.12 and 5.1.13 present typical curves. Table 5.1.10 moves the excitation parameters farther away from the origin. Figure 5.1.14 gives a representative set of plots, and Figure 5.1.15 is a blow up of Figure 5.1.14 to give a better look at the reduced model performance. Tables 5.1.11 and 5.1.2 are tables with constant initial state conditions and randomly chosen frequency and amplitude pairs. Table 5.1.11 has amplitudes in the range (-0.05, 0.05), and Table 5.1.12 has amplitudes in the range (-0.25, 0.25). Both have frequencies chosen from the range (2,6). Finally, Table 5.1.13 tests the low frequency and d.c. behavior of the full and reduced second degree models.

From all of the results presented so far it appears the model reduction technique is working quite well as both criteria for judging reduced models

1 0.001 0.001 0.000 0.000 0.75 1.00 0.713 2 0.001 0.001 0.050 0.050 0.75 1.00 0.765 3 0.001 0.001 0.050 -0.050 0.75 1.00 0.278	
4 0.001 0.001 -0.050 -0.050 0.75 1.00 0.486 5 0.001 0.001 -0.050 0.050 0.75 1.00 0.240 6 0.001 0.001 0.150 0.75 1.00 0.523 7 0.001 0.001 0.150 -0.150 0.75 1.00 0.661 8 0.001 0.001 -0.150 0.150 0.75 1.00 0.582 9 0.001 0.001 -0.150 0.150 0.75 1.00 0.582 9 0.001 0.001 0.000 0.000 0.75 1.00 1.15 10 0.001 -0.001 0.050 0.050 0.75 1.00 0.554 12 0.001 -0.001 0.050 0.050 0.75 1.00 0.246 14 0.001 -0.001 0.050 0.75 1.00 0.251 15 0.001 -0.001 0.150 <t< td=""><td>0.837 0.4741 0.4649 0.4</td></t<>	0.837 0.4741 0.4649 0.4

Table 5.1.9 Simulation Table for Second Degree Full Model versus Second Degree Reduced Model



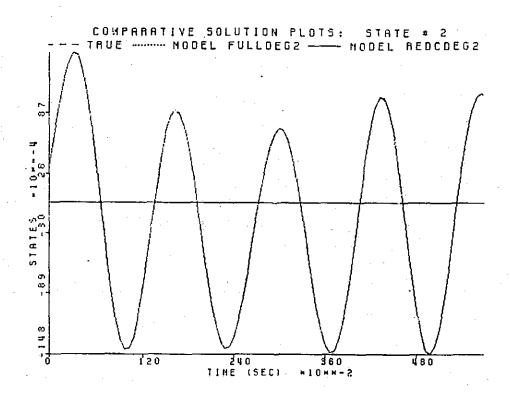
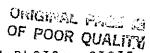


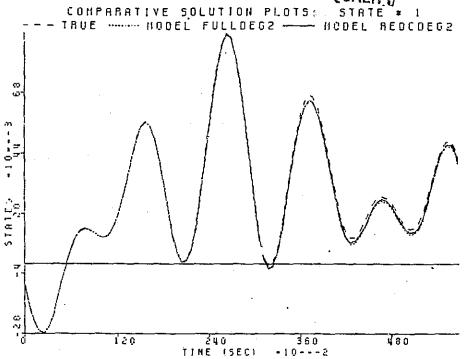
Figure 5.1.12 Simulation Number 32 of Table 5.1.9

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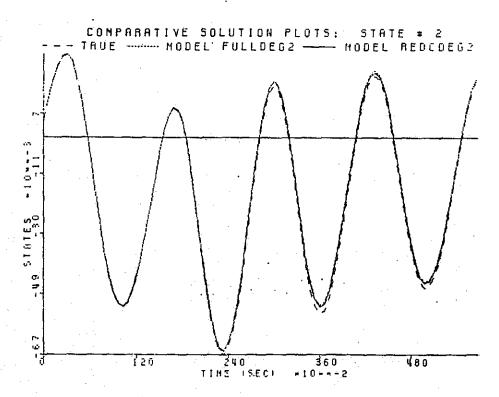
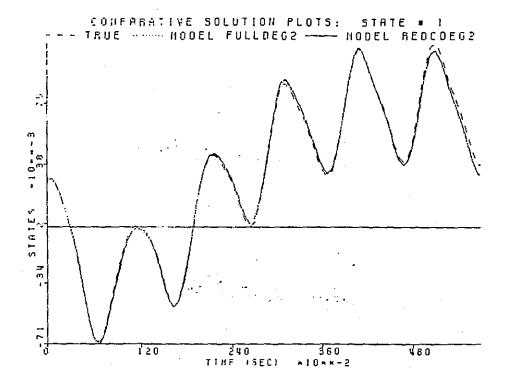


Figure 5.1.13 Simulation Number 26 of Table 5.1.9

Table 5.1.10 Simulation Tables for Second Degree Full Model versus Second Degree Reduced Model



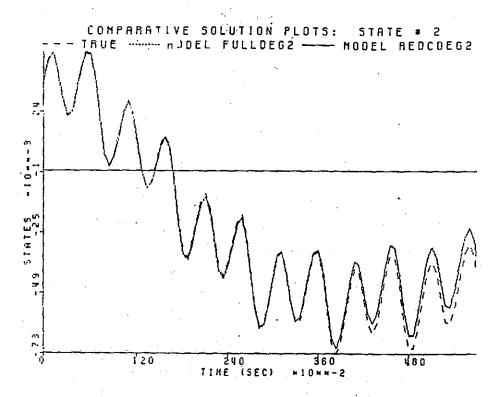
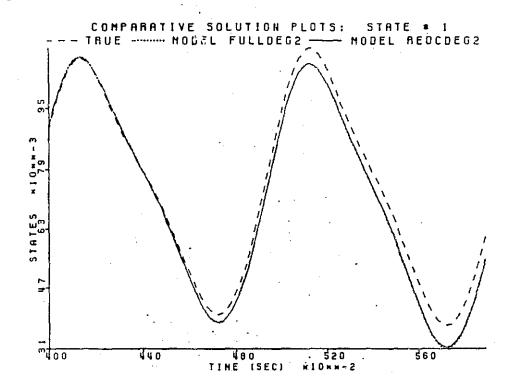


Figure 5.1.14 Simulation Number 19 of Table 5.1.10



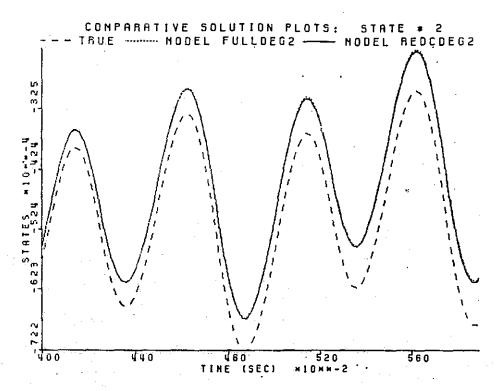
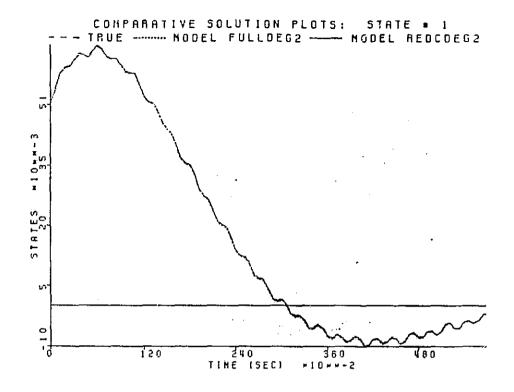


Figure 5.1.15 Expanded Section of Figure 5.1.14

original page is of poor quality

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1 0.050	-0.032 -0.049 0.004 0.018 0.043 -0.019 0.027 -0.021 0.027 -0.022 -0.027 -0.022 -0.027 -0.025 -0.027 -0.027 -0.027 -0.027 -0.041 -0.027 -0.045 -0.027 -0.027 -0.035 -0.027 -0.035 -0.027 -0.035 -0.027 -0.035 -0.027 -0.036 -0.027 -0.036 -0.027 -0.038 -0.027 -0.039 -0.027 -0.039 -0.027 -0.039 -0.027 -0.028 -0.027 -0.039 -0.029	873201839799957122051238781193010225941303492831447 34323324342342255542234234234245544522233432534325343225 2404511019630821294147900472998214082309140331447 524342555323335542323332342255243333432345542553	0.140 0.320E-01 0.898E-02 0.325 0.238 0.224 0.165 0.104 0.819E-01 0.409E-01 0.104 0.819E-01 0.527E-01 0.284E-01 0.527E-01 0.103 0.227 0.211 0.227 0.325E-01 0.934E-02 0.325E-01 0.934E-02 0.325E-01 0.934E-02 0.325E-01 0.435E-01 0.435E-01 0.455E-01 0.477P 0.327E-01 0.477P 0.327E-01 0.243 0.387E-01 0.179 0.327E-01 0.463E-01 0.179 0.327E-01 0.389E-01 0.179 0.327E-01 0.145E-01 0.179 0.327E-01 0.145E-01 0.179 0.327E-01 0.145E-01 0.179 0.327E-01 0.147E-01 0.258 0.244 0.626E-01 0.385E-01 0.147E-01 0.463E-01 0.160E-01 0.404E-01 0.160E-01 0.404E-01 0.160E-01 0.258 0.224 0.207 0.325E-01 0.2480 0.250 0.276 0.2680 0.278 0.2680 0.278 0.2680 0.278 0.2680 0.278 0.2680 0.278 0.2680 0.278 0.2680 0.278 0.2680 0.278 0.278 0.370E-01 0.134 0.134 0.486E-01 0.136 0.134

Table 5.1.11 Simulation Table for Second Degree Full Model versus Second Degree Reduced Model



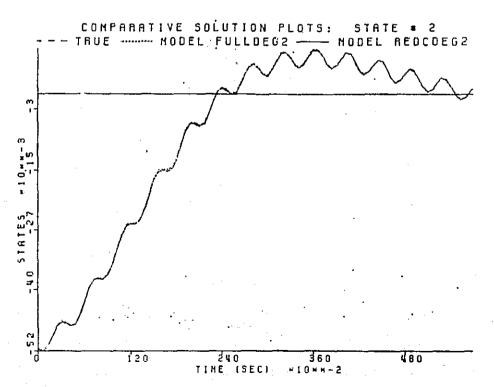
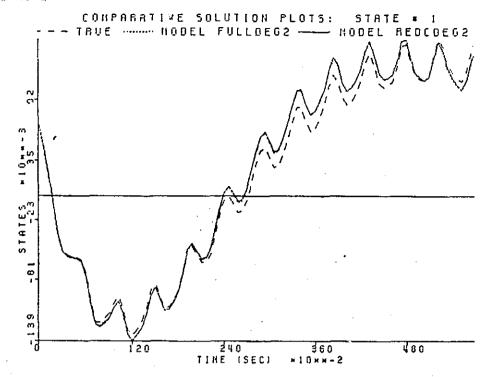


Figure 5.1.16 Simulation Number 13 of Table 5.1.11

S ‡	INITIAL CON	DITIONS	AMPLI	TUDES	FREQUE	NCIES	ERROR	RATIOS
12345678901234567890123456789012345678901234567890	0.0755555555555555555555555555555555555	0.0755555555555555555555555555555555555	01158760554455909765007914655464992913881370447666061101843554455909765007914655464992913881370447666061121798666666666666666666666666666666666666	0.1192420216085504642634495481396350592306574856107888139635059233065748561078888550100000000000000000000000000000	409241912152972743484334334330349371257045446710514 45323342543254343355542544443532323554455553391723224	20704507275403667884775071447318324708523603108790 2070450722544425535423542245454545252234222544434254525	1211121111311133312111211311111121213111111	311112463679865974737124720205465210421506656404575 99.6122364247007386009239725210647025553551004260201 0311211161154612113181111121413461111111515151111111111111111111111

Table 5.1.12 Simulation Table for Second Degree Full Model versus Second Degree Reduced Model



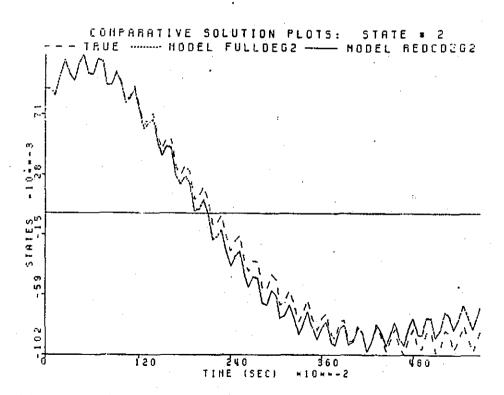
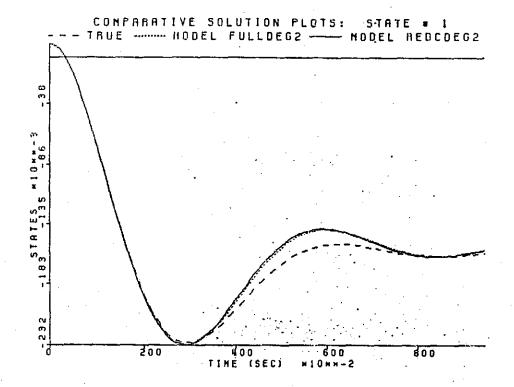


Figure 5.1.17 Simulation Number 1 of Table 5.1.1

1 0.001 -0.001 0.010 -0.075 0.00 0.00 2 0.001 -0.001 0.010 -0.075 0.01 0.01	0.312 0.318 0.300 0.307 0.260 0.27
2 0.001 -0.001 0.010 -0.075 0.02 0.02 4 0.001 -0.001 0.010 -0.075 0.02 0.02 5 0.001 -0.001 -0.050 0.050 0.00 0.00 6 0.001 -0.001 -0.050 0.050 0.01 0.01 7 0.001 -0.001 -0.050 0.050 0.02 0.02 8 0.001 -0.001 -0.050 0.050 0.05 0.05 9 0.001 -0.001 -0.075 0.010 0.00 0.00 10 0.01 -0.001 -0.075 0.010 0.01 0.01 11 0.001 -0.001 -0.075 0.010 0.02 0.02 12 0.001 -0.001 -0.075 0.010 0.05 0.05 13 0.001 -0.001 -0.075 0.010 0.05 0.05 14 0.01 -0.001 -0.075 0.010 0.05 0.05 15 0.01 -0.001 -0.075 0.075 0.01 0.01 15 0.01 -0.001 -0.075 -0.075 0.01 0.01 15 0.01 -0.001 -0.075 -0.075 0.01 0.01 17 0.010 -0.010 0.010 -0.075 0.05 17 0.	0.210 0.232 1.67 1.80 1.73 1.80 1.73 1.20 1.23 1.20 1.34 1.34 1.34 1.81 0.694 0.7183 0.7183 0.2298 0.2298 0.2298 0.2298 0.2298 0.2298 0.2298 1.72 1.72 1.72 1.72 1.72 1.73 1.72 1.73 0.74 0.75 0.7

Table 5.1.13 Low Frequency Table for Second Degree Full Model versus Second Degree Reduced Model

OHOMAL POSTANIA



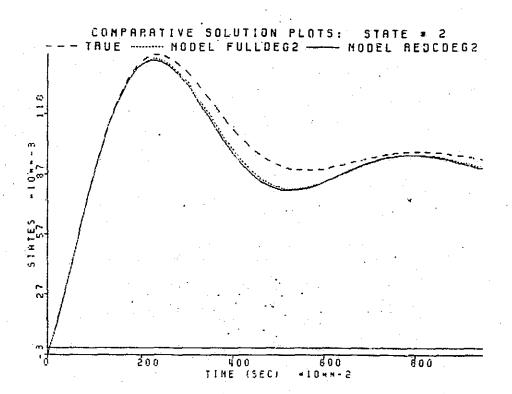


Figure 5.1.18 Simulation Number 26 of Table 5.1.13

were met. But, in this case the full model has only fourteen terms which is not unreasonably large. The true usefulness of the reduction method will appear when used on models with many nonlinear terms. With this in mind, the third degree approximation of this same two state system is studied.

5.1.2 DEGREE THREE APPROXIMATION

In this example, the third degree approximation now has 34 terms corresponding to the products

x1, x2, u1, u2, x1x1, x1x2, x2x2,

x1u1, x1u2, x2u1, x2u2, u1u1, u1u2,

u2u2, x1x1x1, x1x1x2, x1x2x2,

x2x2x2, x1x1u1, x1x1u2, x1x2u1, x1x2u2,

x2x2u1, x2x2u2, x1u1u1, x1u1u2, x1u2u2,

x2u1u1, x2u1u2, x2u2u2, u1u1u1, u1u1u2,

u1u2u2, u2u2u2.

The appendix contains the locally optimum, third degree model to be used as the starting model for the reduction. See Figure 5.1.19 for the original third degree model and the "first pass" reduction test. Note the model is identified using initial state conditions (0.001, -0.001). It also uses as control input excitations the same sum of two sinusoids, although with different amplitudes and frequencies,

 $u_1(t) = 0.055 \sin(2\pi t \cdot \phi_1) \div 0.055 \sin(2\pi t \cdot 2\phi_1)$ $u_2(t) = 0.0325 \sin(2\pi t \cdot \phi_2) + 0.0325 \sin(2\pi t \cdot 2\phi_2)$ where $\phi_1 = 0.5090$ and $\phi_2 = 2.0098$.

Directing our attention to the reduction test and the terms with changes in errors of magnitude less than 4, it is found that 18 terms can be omitted. The remaining 16 terms are

ENTER INITIAL CONDITIONS FOR THE 2 STATES: 0.0010 -0:0010

ENTER SAMPLING FARAMETERS:
APPROXIMATION DEGREE
\$ SAMPLE POINTS

INTERVAL BETWEEN SAMPLES

INTEGRATION STEPSIZE 100 0.0400 0.0050

SHOULD THE EXCITATION BE SINUSOIDAL OR COSINUSOIDAL?[5/C]:

IDENTIFICATION WILL BE DONE WITH SINES.

ENTER THE NUMBER OF (CO)SINUSCIDS FER INPUT:

ENTER THE 2 FREQUENCY WEIGHTS: 0.5090 2.0098

SHOULD THE IDENTIFICATION BE PERFORMED USING DATA THAT IS REDUCED?(Y/N):

ENTER OFTIMIZATION OPTION:

1 - NO OFTIMIZATIONS
2 - AMPLITUDES ONLY
3 - FREQUENCIES ONLY
4 - INITIAL CONDITIONS ONLY
5 - AMPLITUDES AND FREQUENCIES
6 - AMPLITUDES AND INITIAL CONDITIONS
7 - FREQUENCIES AND INITIAL CONDITIONS
8 - AMPLITUDES, FREQUENCIES, AND INITIAL CONDITIONS

0

DO YOU WISH TO NORMALIZE THE DATA?[Y/N]:

THE MATRIX OF SAMPLED MONOMIAL TERMS 34 ROWS AND 100 COLUMNS.

NUMBER OF TIMES COST FUNCTION WAS EVALUATED:

WOULD YOU LIKE TO SEE PLOTS OF THE INPUTS? IF SO, LINE UP THE CARRIAGE.

TSO SPEAKEASY III PI+ 9:57 PM FEBRUARY 22, 1984 :_SIZE=500;GET IDENT;IDENT;QUIT

EXECUTION STARTED

PARTITION NUMBER 1. PRIITION (A 2 BY 2 ARRAY)

Figure 5.1.19a Full Third Degree Model and Reduction Test (First Pass)

```
CRICKY CONTROL
```

```
-1.9977 -3.0008
                       1.0002
  ...WITH EIGENVALUES;
VALUES (A VECTOR WITH 2 COMPONENTS)
-.49874+.86767i -.49874-.86767i
 PARTITION NUMBER 2,
PRTITION (A 2 BY 2 ARRAY)
.0019891 .99779
-1.0001 -6.2231E-7
 PARTITION NUMBER 3.

PRTITION (A 2 BY 3 ARRAY)

-.19024 -.012365 -.0034771

.036033 .0022184 1.5017E-4
 FARTITION NUMBER 4.

PRITITION (A 2 BY 4 ARRAY)

-4.0692 .026129 .0279

.021083 .002591 .0026
                                                       .027909 -.0078805
.0026825 .0035398
 PARTITION NUMBER 5.
PRIITION (A 2 BY 3 ARRAY)
-.0039675 .0071601 -.012386
-.99787 .0010537 7.4681E-4
 PARTITION NUMBER 6.
PRITITION (A 2 BY 4 ARRAY)
.70548 1.6627 3.3163 .78087
-.2223 -.6684 -1.2624 -.3395
 PARTITION NUMBER 7.
   PRTITION (A 2 BY 6 ARRAY)
.62148 -.61366 -.25223
-.28909 -.025264 -.05473
                                                                           -.76455 .44724 -.67544
-.058202 -.048646 -.0056896
PARTITION NUMBER 8.
PRIITION (A 2 BY 6 ARRAY)
-3.5488 -.10235 .41411 .16518 .036896 .075615
-.17155 .96002 -.044252 -.055923 -.035393 -.013022
PARTITION NUMBER 9.
PRIITION (A 2 BY 4 ARRAY)
-.18738 -.032069 -.0251 -.019747
-.51749 -.016328 -.019048 -.0011802
     S (A 34 COMPONENT ARRAY)
12.581 8.5967 7.997
2.6574 2.5454 2.1057
1.1668 .99995 .92334
.40492 .35004 .32218
                                                                        6.0242
1.8416
.82698
.20171
                                                                                                                     4.1246
1.4722
.71888
                                                                                              4.3769
1.5116
.76732
.18144
                                                                                                                                            3.1931
1.3897
.5978
                                                                                                                                                                   2.9288
1.3138
.53451
                                                                                                                      .11826
                                                                                                                                                                     .045849
.084429 .063016 .2218 .201/1
THE MAXIMUM SINGULAR VALUE:
MAX = 12.581
THE MINIMUM NONZERO SINGULAR VALUE:
MIN = .023934
RTIO = 525.67
```

Figure 5.1.19b Full Third Degree Model and Reduction Test (First Pass)

WANT TO TRY REDUCTION TEST?[Y/N3: Y

FULL SYSTEM ERROR: 200 FULL STATE ERRORS: 153	3.5798 2.0251	56.5547
TERM: X1. COLUM CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	7 ‡: 1. 273.2820 226.0011	47,2809
TERM: X2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	1 #: 2. -18.8934 -16.9980	-1.8954
TERM: U1. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	N #: 3. -25.4317 0.0450	-25.4768
TERM: U2. COLUMI CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	V #: 4, 14.0826 14.0826	-0.0000
TERM: X1.X1. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	12.2573 11.6348	0.6225
TERM: X1.X2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	V # 1 6.	
TERM: X2.X2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	7. 0.2715 0.2671	0.0044
TERM: X1.U1. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	1 #: 8.	
TERM: X1.U2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	/ #: 9. -0.0250 -0.0373	0.0123
YERM: X2.U1. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	10. -0.3615 -0.3495	-0.0120
TERM: X2,U2, COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	11. -0.0061 -0.0176	0.0114
TERM: U1.U1. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:		
TERM: U1.U2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	1 4.1 17.	-0.0119
TERM: U2.U2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	2 #: 14. 0.3772	0.0173
TERM: X1.X1.X1. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	1 *: 15. -29.9081 -27.9561	-1.9520

Figure 5.1.19c Full Third Degree Model and Reduction Test (First Pass)

ORIGINAL PACE IN OF POOR QUALITY

TERM: X1.X1.X2. COLUMN	1: 16.	OF POOR QUALIT
CHANGE IN SYSTEM ERRORS: CHANGE IN STATE ERRORS:	135,7215 122,6539	14.9676
TERM: X1.X2.X2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	#: 17. -77.4448 -75.6628	-1.7820
TERM: X2.X2.X2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	#: 18. 79.0131 69.5179	7.495 2
TERM: X1.X1.U1. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	19. 7.9639 7.8924	0.0715
TERM: X1.X1.U2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	#: 20. 1.3980 1.5089	-0.1108
TERM: X1.X2.U1. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	#: 21. 1.7702 1.6826	0.0876
TERM: X1.X2.U2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	#: 22. 2.6437 2.5047	0.1390
TERM: X2.X2.U1. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	4: 93.	-0.3789
TERM: X2.X2.U2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	#: 24. 5.6993 5.7065	-0.0071
TERM: X1.U1.U1. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	\$: 25. 57.3052 58.0436	-0.7384
TERM: X1.U1.U2. COLTAN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	26. -2.3094 0.0114	-2.3208
TERM: X1.U2.U2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	#: 27, -8.9582 -8.6700	-0.2882
TERM: X2.U1.U1. CGLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	#: 28. 1.1869 0.7543	0.4325
TERM: X2.U1.U2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	#: 29. -0.0037 0.2052	-0.2090
TERM: X2.U2.U2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	\$: 30. 2.7262 2.5737	0.1524
TERM: U1.U1.U1. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	-20.7670	-18.5142
TERM: U1.U1.U2. COLUMN CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:	#: 32. 0.2494 0.2183	0.0311
		•

Figure 5.1.19d Full Third Degree Model and Reduction Test (First Pass)

TERM: U1.U2.U2. COLUMN II 33. CHANGE IN SYSTEM ERROR: -0.3957 CHANGE IN STATE ERRORS: -0.2063 -0.1894

COLUMN #: 34.

TERM: U2.U2.U2. COLU CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS: 0.2591 0.2608 -0.0018

DO YOU WANT TO RESPARD ANY TERMS AND RE-OPTIMIZETEY/NO: Y

HOW MANY TERMS WILL BE KEPT? 16

ENTER THE COLUMN NUMBERS OF COLUMNS WHICH ARE TO BE KEPT. ENTRY OF COLUMN NUMBERS WILL TAKE PLACE IN BLOCKS OF 10. WHEN ASKED, ENTER A AS AN ARRAY WITH A MAXIMUM OF TEN ELEMENTS. A = ARRAY(10:1,2,3,4,5,8,12,15,16,17)

A = ARRAY(6:18,19,24,25,27,31)

MANUAL MODE SPACE USED 113 K NOW, 138 K PEAK, SIZE 500 K DO YOU WISH TO SAVE THIS MODEL?CY/NJ:N DO YOU WISH TO IDENTIFY ANOTHER MODEL?CY/NJ;N

Figure 5.1.19e Full Third Degree Model and Reduction Test (First Pass)

Constant British Will THE PEOR CUALITY

x1, x2, u1, u2, x1x1, x1u1, u1u1, x1x1x1, x1x1x2, x1x2x2, x2x2x2, x1x1u1, x2x2u2, x1u1u1, x1u2u2, u1u1u1.

These terms are kept and a new local optimum is found. Optimization over frequencies yielded $\phi_1 = 0.3950$ and $\phi_2 = 1.3506$. Sex Figure 5.1.20 for the "second pass". Note that the IMSL optimization routine ended with a "terminal error". This error (IER = 130) is an IMSL provided error check to insure that roundoff error does not become excessive. Note further that a FORTRAN error message summary is provided. In this instance error 208 was encountered 34 times. Error 208 is an underflow error and indicates that an internal variable was less than 10^{-78} . The standard fixup is to set that variable to zero, which is perfectly acceptable.

The "second pass" reduction test indicates that another 3 terms are fairly insignificant. This brings the total number of terms to 13. They are x1, x2, u1, u2, x1u1, u1u1, x1x1x1, x1x1x2, x2x2x2, x1x1u1, x2x2u2, x1u1u1, u1u1u1.

Again, only these 13 terms are kept and a locally optimum frequency set is determined. The model is identified and a reduction test performed. (See Figure 5.1.21 for this "third pass".) The reduction test indicates that term 19 (x1x1u1) could be removed, but after execution of the "fourth pass", the model simulations showed that the 12 term model was not acceptable. So the 13 term model is the final model, and the final reduction is 21 terms removed out of a possible 34; a 62% reduction.

```
ENTER INITIAL COMDITIONS FOR THE 2 STATES!
                                     -0.0010
         ENTER SAMPLING PARAMETERS:
APPROXIMATION DEGREE
* SAMPLE POINTS
                           INTERVAL BETWEEN SAMPLES
INTEGRATION STEPSIZE
0.0400 0.0050
                    100
         SHOULD THE EXCITATION BE SINUSOIDAL OR COSINUSOIDAL?CS/CJ:
         IDENTIFICATION WILL BE DONE WITH SINES.
         ENTER THE NUMBER OF (CO)SINUSOIDS PER INPUT:
                              4 INITIAL INPUT AMPLITUDES:
          0.550000000000000002E-01 0.32500000000000002E-01
         ENTER THE 2 FREQUENCY WEIGHTS: 0.5090 2.9098
         SHOULD THE IDENTIFICATION BE PERFORMED USING DATA THAT IS REDUCED?[Y/N];
         ENTER OFTIMIZATION OFTION:

1 - NO OFTIMIZATIONS
2 - AMPLITUDES ONLY
3 - FREQUENCIES ONLY
4 - INITIAL CONDITIONS ONLY
5 - AMPLITUDES AND FREQUENCIES
6 - AMPLITUDES AND INITIAL CONDITIONS
7 - FREQUENCIES AND INITIAL CONDITIONS
8 - AMPLITUDES, FREQUENCIES, AND INITIAL CONDITIONS
         ENTER THE TRACE OF THE COVARIANCE MATRIX: 0.99999999999999999954E-06
         ENTER THE OPTIMIZATION METHOD: 1,2,3 OR 4
         ENTER # OF SIGNIFICANT DIGITS FOR CONVERGENCE, MAXIMUM # OF FUNCTION CALLS, AND THE ZXMIN OPTION # (0,1,2,3)
2 200 2
         10 CALLS OF FCN....
         0.52531470+00 0.2009603D+01
THE CONDITION NUMBER: 0.152D+10
OBJECTIVE FUNCTION VALUE: 0.5532D+04
        20 CALLS OF FCN....
THE PARAMETERS:
0.5254612D+00 0.2008699D+01
THE CONDITION NUMBER: 0.152D+10
DBJECTIVE FUNCTION VALUE: 0.5516D+04
          30 CALLS OF FCN....
        THE FARAMETERS:
0.5199821D+00 0.1989839D+01
THE CONDITION NUMBER: 0.145D+10
OBJECTIVE FUNCTION VALUE: 0.5316D+04
Figure 5.1.20a Reduced Third Degree Model and Reduction Test (Second Pass)
```

40 CALLS OF FCN....
THE PARAMETERS:
0.5039734D+00 0.1890305D+01
THE CONDITION NUMBER: 0.123D+10
OBJECTIVE FUNCTION VALUE: 0.4721D+04

50 CALLS OF FCN....
THE PARAMETERS:
0.4076332D+00 0.1414816D+01
THE CONDITION NUMBER: 0.470D+09
0BJECTIVE FUNCTION VALUE: 0.2133D+04
*** TERMINAL ERROR - (IER = 130) FROM IMSL ROUTINE ZXMIN

DO YOU WISH TO NORMALIZE THE DATA?[Y/N]:

THE MATRIX OF SAMPLED MONOMIAL TERMS HAS 16 ROWS AND 100 COLUMNS.

NUMBER OF TIMES COST FUNCTION WAS EVALUATED:

52

THE OPTIMUM FREQUENCIES ARE: FREQ(1)= 0.395 FREQ(2)= 1.351 FREQ(3)= 0.000

WOULD YOU LIKE TO SEE PLOTS OF THE INPUTS? IF SO, LINE UP THE CARRIAGE.

MESSAGE SUMMARY: MESSAGE NUMBER - COUNT 208 34

TSO SPEAKEASY III FI+ 10:59 PM FEBRUARY 22, 1984 :_SIZE=500;GET IDENT;IDENT;QUIT

EXECUTION STARTED COLSKEPT (A 16 COMPONENT ARRAY) 1 2 3 4 5 8 12 15 16 17 18 19 24 25 27 31

FARTITION NUMBER 1. PRIITION (A 2 BY 2 ARRAY) -1.9999 -3.0002 1.0011 .99978

...WITH EIGENVALUES: VALUES (A VECTOR WITH 2 COMPONENTS) -.50005+.86841i -.50005-.86841i

PARTITION NUMBER 2. PRIITION (A 2 BY 2 ARRAY) 6.5116E-4 .99902 -.99844 -.0013877

PARTITION NUMBER 3.
PRITITION (A 2 BY 3 ARRAY)
-.028226 0 0
-.093347 0 0

Figure 5.1.20b Reduced Third Degree Model and Reduction Test (Second Pass)

```
FARTITION NUMBER 4.
FRTITION (A 2 BY 4 ARRAY)
-4.0367 0 0
-.062668 0 0
PARTITION NUMBER 5.
PRIITION (A 2 BY 3 ARRAY)
-.0022205 0 0
-1.0093 0 0
FARTITION NUMBER 6.
FRTITION (A 2 BY 4 ARRAY)
-.91373 .39023 .40494 -.40803
.48991 -1.1593 -.11776 .55633
PARTITION NUMBER 7.
PRITITION (A 2 BY 6 ARRAY)
.11824 0 0
                                                                                                    .049393.
    .58555
                        0
                                                                                                     .62134
FARTITION NUMBER 8.
PRIITION (A 2 BY 6 ARRAY)
-3.9208 0 .028512
.48114 0 .12065
FARTITION NUMBER 9.
FRTITION (A 2 BY 4 ARRAY)
-.058225 0 0
-.57913 0 0
    S (A 16 COMPONENT ARRAY)
11.449 7.1423 6.0617
1.1688 1.1436 .8219
                                                             5.0361
                                                                                 2.7425
                                                                                                    2.088
                                                                                                                       1.9426
                                                                                                                                          1.3796
                                                                                 .31614
                                                                                                                        ·084745
THE MAXIMUM SINGULAR VALUE:

MAX = 11.449

THE MINIMUM NONZERO SINGULAR VALUE:

MIN = .034909
...AND THEIR RATIO: RTIO = 327.96
WANT TO TRY REDUCTION TEST?[Y/N]: Y
FULL SYSTEM EPROR: FULL STATE ERRORS:
                                              138,4610
70,3743
                                                                            68.0666
TERM: X1. COLUMN #:
CHANGE IN SYSTEM ERROR: 19
CHANGE IN STATE ERRORS: 19
TERM: X2. COLUMN #: 2. CHANGE IN SYSTEM ERROR: 234.7848 CHANGE IN STATE ERRORS: 194.0638
TERM: U1. COLUMN $: 3. CHANGE IN SYSTEM ERROR: -41.7820 CHANGE IN STATE ERRORS: -0.0060
                                                                                     -41.7761
TERM: U2. COLUMN #: CHANGE IN SYSTEM ERROR: -
                                                         -16.5398
```

Figure 5.1.20c Reduced Third Degree Model and Reduction Test (Second Pass)

CHANGE	IN STATE	ERRORS:	-14,5402	0.0204
TERM: CHANGE CHANGE	IN SYSTEM IN STATE	COLUMN ERROR: ERRORS:	1: 5. 3.0913 0.0624	3.0289
TERM: CHANGE CHANGE	IN SYSTEM IN STATE	COLUMN ERROR: ERRORS:	#: 8. 113.9193 114.6507	-0.7314
TERM: CHANGE CHANGE	IN SYSTER IN STATE	COLUMN FRROR: ERRORS:	#: 12. -19.9546 -0.0270	-19.9276
TERM:	X1.X1.X1	. COLUMN	#: 15.	~10.6505
CHANGE	IN SYSTEM	I ERROR:	4.4767	
CHANGE	IN STATE	ERRORS:	15.1272	
TERM: CHANGE CHANGE	IN SYSTEM	. COLUMN ERROR: ERRORS:	\$: 16. -2.5463 5.6283	-8.1746
TERM:	X1.X2.X2	COLUMN	#: 17.	2.1641
CHANGE	IN SYSTEM	ERROR:	-1.4812	
CHANGE	IN STATE	ERRORS:	-3.6453	
TERM:	X2.X2.X2	COLUMN	#: 18.	10.0051
CHANGE	IN SYSTEM	ERROR:	2.7722	
CHANGE	IN STATE	ERRORS:	-7.2330	
TERM:	X1.X1.U1	. COLUMN	#: 19.	7,7985
CHANGE	IN SYSTEM	I ERROR:	7.1912	
CHANGE	IN STATE	ERRORS:	-0.6173	
TERM: CHANGE CHANGE	IN SYSTEM	. COLUMN ERROR: ERRORS:	~6.0369	-5.1057
TERM:	X1.U1.U1	. COLUMN	#: 25.	-2.4901
CHANGE	IN SYSTEM	ERROR:	9.4809	
CHANGE	IN STATE	ERRORS:	11.9710	
TERM:	X1.U2.U2	. COLUMN	#: 27.	-1.3695
CHANGE	IN SYSTEM	ERROR:	-1.2075	
CHANGE	IN STATE	ERRORS:	0.1620	
TERM:	U1.U1.U1	COLUMN	#: 31.	-25.5545
CHANGE	IN SYSTEM	ERROR:	-24.7955	
CHANGE	IN STATE	ERRORS:	0.7590	

DO YOU WANT TO DISCARD ANY TERMS AND RE-OPTIMIZE?[Y/N]: Y
HOW MANY TERMS WILL BE KEPT? 13

ENTER THE COLUMN NUMBERS OF COLUMNS WHICH ARE TO BE KEPT.
ENTRY OF COLUMN NUMBERS WILL TAKE PLACE IN BLOCKS OF 10.
WHEN ASKED, ENTER A AS AN ARRAY WITH A MAXIMUM OF TEN ELEMENTS.
A = ARRAY(1,2,3,4,8,12,15,16,18,19)

A = ARRAY(24,25,31)

MANUAL MODE
SPACE USED 70 K NOW, 81 K PEAK, SIZE 500 K
DO YOU WISH TO SAVE THIS MODEL?[Y/N]:N
DO YOU WISH TO IDENTIFY ANOTHER MODEL?[Y/N]:N

Figure 5.1.20d Reduced Third Degree Model and Reduction Test (Second Pass)

```
ENTER INITIAL CONDITIONS FOR THE 2 STATES;
         0,0010
                            -0,0010
ENTER SAMPLING PARAMETERS:
                  APPROXIMATION DEGREE
                  # SAMPLE POINTS
INTERVAL BETWEEN SAMPLES
INTEGRATION STEPSIZE
          100
                       0.0400
                                         0.0050
SHOULD THE EXCITATION BE SINUSOIDAL OR COSINUSOIDAL?[5/C]:
IDENTIFICATION WILL BE DONE WITH SINES.
ENTER THE NUMBER OF (CO)SINUSOIDS FER INPUT:
 ENTER THE 4 INITIAL INPUT AMPLITUDES: .0.550000000000000002E-01
ENTER THE 2 FREQUENCY WEIGHTS: 0.3950 1.3506
SHOULD THE IDENTIFICATION BE PERFORMED USING DATA THAT IS REDUCED?CY/NJ:
ENTER OPTIMIZATION OFTION:
                IMIZATION OFTINIS

1 - NO OFTIMIZATIONS

2 - AMPLITUDES ONLY

3 - FREQUENCIES ONLY

4 - INITIAL CONDITIONS ONLY

5 - AMPLITUDES AND FREQUENCIES

6 - AMPLITUDES AND INITIAL CONDITIONS

7 - FREQUENCIES AND INITIAL CONDITIONS

AMELITUDES - EREQUENCIES, AND INITIAL
                   - AMPLITUDES, FREQUENCIES, AND INITIAL CONDITIONS
ENTER THE TRACE OF THE COVARIANCE MATRIX: 0.99999999999999954E-06
ENTER THE OPTIMIZATION METHOD: 1,2,3 OR 4
ENTER # OF SIGNIFICANT DIGITS FOR CONVERGENCE, MAXIMUM # OF FUNCTION CALLS, AND THE ZXMIN OFTION # (0,1,2,3)
           200
10 CALLS OF FCN....
0.3986263D+00 0.1343124D+01
THE CONDITION NUMBER: 0.400D+08
OBJECTIVE FUNCTION VALUE: 0.1685D+03
20 CALLS OF FCN....
THE PARAMETERS:
0.39918380+00 0.13521940+01
THE CONDITION NUMBER: 0.3970+08
OBJECTIVE FUNCTION VALUE: 0.16810+03
30 CALLS OF FCN....
THE FARAMETERS:
0,39916420+00 0.13521550+01
THE CONDITION NUMBER: 0.3970+08
OBJECTIVE FUNCTION VALUE: 0.16810+03
```

Figure 5.1.21a Reduced Third Degree Model and Reduction Test (Third Pass)

40 CALLS OF FCN....
THE PARAMETERS:
0.3991505D+00 0.1352110D+01
THE CONDITION NUMBER: 0.397D+08
OBJECTIVE FUNCTION VALUE: 0.1681D+03

CONVERGENCE WAS ACHIEVED AND NO ERRORS OCCURRED.

DO YOU WISH TO NORMALIZE THE DATA?EY/NJ:

THE MATRIX OF SAMPLED MONOMIAL TERMS HAS 13 ROWS AND 100 COLUMNS.

NUMBER OF TIMES COST FUNCTION WAS EVALUATED:

THE OPTIMUM FREQUENCIES ARE: FREQ(1)= 0.399 FREQ(2)= 1.352 FREQ(3)= 0.000

48 .

WOULD YOU LIKE TO SEE PLOTS OF THE INPUTS? IF SO, LINE UP THE CARRIAGE. N

MESSAGE SUMMARY: MESSAGE NUMBER - COUNT 208 55

TSO SPEAKEASY III PI+ 11:30 PM FEBRUARY 22, 1784 :_SIZE=500;GET IDENT;IDENT;QUIT

EXECUTION STARTED COLSKEPT (A 13 COMPONENT ARRAY) 1 2 3 4 8 12 15 16 18 19 24 25 31

FARTITION NUMBER 1. PRITITION (A 2 BY 2 ARRAY) -1.9999 -2.9999 .99957 .99945

...WITH EIGENVALUES:
VALUES (A VECTOR WITH 2 COMPONENTS)
-.50021+.86582i -.50021-.86582i

FARTITION NUMBER 2. FRTITION (A 2 RY 2 ARRAY) 6.3827E-4 .99898 -.99849 -.0014354

PARTITION NUMBER 3. FRIITION (A 2 RY 3 ARRAY) 0 0 0 0 0 0

PARTITION NUMBER 4.
PRITITION (A 2 BY 4 ARRAY)
-4.0226 0 0 0
-.026954 0 0 0

Figure 5.1.21b Reduced Third Degree Model and Reduction Test (Third Pass)

C - 2

```
PARTITION NUMBER 5.
PRITITION (A 2 BY 3 ARRAY)
-6.5226E-4 0 0 0
-1.0026 0 0
FARTITION NUMBER 6.
FRTITION (A 2 BY 4 ARRAY)
-1.2411 .15993 0 -..
-.26768 -.91146 0 -..
                                               -.5407
                                                       .77552
PARTITION NUMBER 7.
PRITITION (A 2 BY 6 ARRAY)
-.063216 0 0
.20884 0 0
                                                                                                    .073678
                                                                                                    .68143
PARTITION NUMBER 8.
PRITITION (A 2 BY 6 ARRAY)
-4.0074 0 0 0
.27416 0 0 0
PARTITION NUMBER 9.
PRTITION (A 2 BY 4 ARRAY)
-.078589 0 0
-.66691 0 0
S (A 13 COMPUNENT ARRAY)

9.8648 7.1115 6.0803 4.9378 2.4718 1.9875 1.3949 1.2823 1.1637

.95869 .56438 .26112 .16092

THE MAXIMUM SINGULAR VALUE:

MAX = 9.8648
 THE MINIMUM NONZERO SINGULAR VALUE: MIN = .16092
...AND THEIR RATIO: RTIO = 61.303
WANT TO TRY REDUCTION TEST?[Y/N3: Y
 FULL SYSTEM ERROR: FULL STATE ERRORS:
 TERM: X1. COLUMN #: 1.
CHANGE IN SYSTEM ERROR: 121.0994
CHANGE IN STATE ERRORS: 85.4754
                                                                                       35.6240
 TERM: X2. COLUMN #: 2.
CHANGE IN SYSTEM ERROR: 222.9544
CHANGE IN STATE ERRORS: 220.0404
                                                                                         2.9140
 TERM: U1. COLUMN *: 3. CHANGE IN SYSTEM ERROR: -37.8873 CHANGE IN STATE ERRORS: -0.0055
                                                                                     -37.8818
 TERM: U2. COLUMN #: 4. CHANGE IN SYSTEM ERROR: -15.6390 CHANGE IN STATE ERRORS: -15.6634
                                                                                         0.0243
                                        COLUMN #: 8.
ROR: 123.9281
 TERM: X1.U1. COLUNCHANGE IN SYSTEM ERROR:
```

Figure 5.1.21c Reduced Third Degree Model and Reduction Test (Third Pass)

```
CHANGE IN STATE ERRORS!
                                                           124,1391
                                                                                          -0.2010
TERM: U1.U1. COLUMN : 12.
CHANGE IN SYSTEM ERROR: -27.3552
CHANGE IN STATE ERRORS: -0.0115
                                                                                        -27.5438
TERM: X1.X1.X1. COLUNN #: 15.
CHANGE IN SYSTEM ERROR: 15.9879
CHANGE IN STATE ERRORS: 15.2055
                                                                                             0.7824
TERM: X1.X1.X2. COLUMN $: 16.
CHANGE IN SYSTEM ERROR: 16.6559
CHANGE IN STATE ERRORS: 0.5196
                                                                                           16,1362
TERM: X2.X2.X2. COLUMN *: 18,
CHANGE IN SYSTEM ERROR: -8.7379
CHANGE IN STATE ERRORS: -4.3799
                                                                                           -4.3580
TERM: X1.X1.U1. COLUMN *: CHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:
                                                               19.
1.9149
0.4854
                                                                                             1,4295
TERM: X2.X2.U2. COLUMN *: 24.
CHANGE IN SYSTEM ERROR: -5.8982
CHANGE IN STATE ERRORS: -1.3748
                                                                                            -4.5234
TERM: X1.U1.U1. COLUMN #: 25.
CHANGE IN SYSTEM ERROR: 7.7140
CHANGE IN STATE ERRORS: 7.7567
                                                                                           -0.0427
TERM: U1.U1.U1. COLUMN *:
CHANGE IN SYSTEM ERROR: -:
CHANGE IN STATE ERRORS:
                                                           11.
-25.8595
0.9593
                                                                                         -26.8188
```

DO YOU WANT TO DISCARD ANY TERMS AND RE-OPTIMIZE?[Y/N]: N

MANUAL MODE
SPACE USED 62 K NOW, 71 K PEAK, SIZE 500 K
DO YOU WISH TO SAVE THIS MODEL?[Y/N]:N
DO YOU WISH TO IDENTIFY ANOTHER MODEL?[Y/N]:N

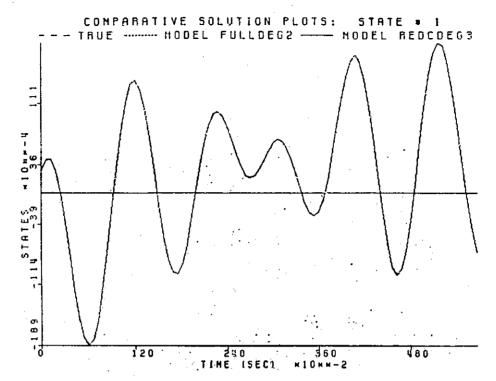
Figure 5.1.21d Reduced Third Degree Model and Reduction Test (Third Pass)

Now all that remains is simulation to bear out the model's validity. First we test against the full second degree model and the error ratio for each state is the mean square of the reduced third degree over the mean square error of the full second degree. In each of the following tables MODEL1 is the second degree model and MODEL2 is the reduced third degree. The first two tables (Table 5.1.14 and Table 5.1.15) and their corresponding plots display the behavior when excitation parameters are chosen close to the origin. The following three tables stretch the initial conditions and amplitudes farther from the origin. From this set of data, it is clear that the reduced third degree model outperforms the full second degree. Table 5.1.19 is a table including randomly chosen amplitudes and frequencies. The corresponding figures (Figures 5.1.29 and 5.1.30) show a worst case and a typical plot of model behavior. The final two tables show the low frequency and d.c. behavior of the two models. Again a definite improvement is observed.

Now, comparison against the full degree three model is in order. The error ratios for each state are now the mean square error of the reduced model over the mean square error of the full third degree approximation. The first table (Table 5.1.22) and its corresponding set of plots (Figure 5.1.34) show the typical comparison of model behavior near the origin. The next two tables move the control amplitudes farther out. Throughout these tables, the reduced model behavior is quite good and sometimes even better than that of the full model, but the reduced model went unstable as the amplitudes reached ± 0.6 , while the full model remained stable and did well (see the appendix). Table 5.1.25 is another random table with amplitudes in the range (-0.4, 0.4) and frequencies in the range (2,6). Table 5.1.26 is a low frequency table and Table 5.1.27 is a table of random steps. In these cases, we note that re-

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1 0.001 0.001 2 0.001 0.001 3 0.001 0.001 4 0.001 0.001 5 0.001 0.001 6 0.001 0.001 7 0.001 0.001 9 0.001 0.001 10 0.001 -0.001 11 0.001 -0.001 12 0.001 -0.001 13 0.001 -0.001 14 0.001 -0.001 15 0.001 -0.001 17 0.001 -0.001 18 0.001 -0.001 19 -0.001 -0.001 17 0.001 -0.001 18 0.001 -0.001 17 0.001 -0.001 20 -0.001 -0.001 21 -0.001 -0.001 22 -0.001 -0.001 23 -0.001 -0.001 24 -0.001 -0.001 25 -0.001 -0.001 27 -0.001 -0.001 28 -0.001 -0.001 29 -0.001 -0.001 30 -0.001 0.001 31 -0.001 0.001 32 -0.001 0.001 33 -0.001 0.001 34 -0.001 0.001 35 -0.001 0.001 36 -0.001 0.001	0.000 0.050 0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.150 -0.150 -0.150 -0.150 -0.150 -0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.150	0.75 1.00 0.00 0.00 0.00 0.00 0.00 0.00 0.0	0.148E-01 0.133E-01 0.226 0.394 0.197 0.419 0.279 0.639 0.614 1.21 0.149 0.947E-01 0.205 0.114 0.141 0.879E-01 0.181 0.105 0.149E-01 0.107E-01 0.321 0.599 0.208 0.415 0.385 0.886 0.705 1.57 0.144 0.929E-01 0.185 0.108 0.134 0.852E-01 0.168 0.101 0.187E-01 0.168E-01 0.323 0.647 0.238 0.472 0.401 0.907 0.950 1.90 0.141 0.919E-01 0.131E-01 0.943E-02 0.141 0.992E-01 0.131E-01 0.943E-02 0.229 0.428 0.222 0.475 0.319 0.737 0.753 1.42 0.147 0.939E-01 0.139 0.871E-01 0.139 0.871E-01 0.139 0.871E-01 0.139 0.871E-01 0.139 0.871E-01 0.139 0.871E-01

Simulation Table for Second Degree Full Model versus Third Degree Reduced Model



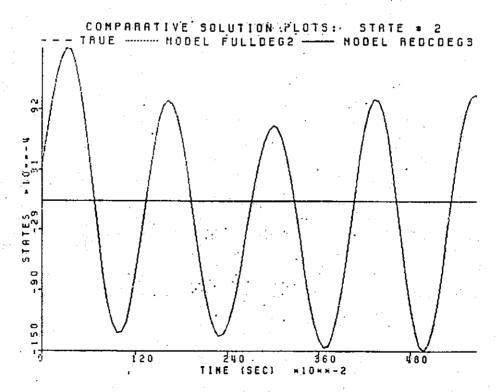
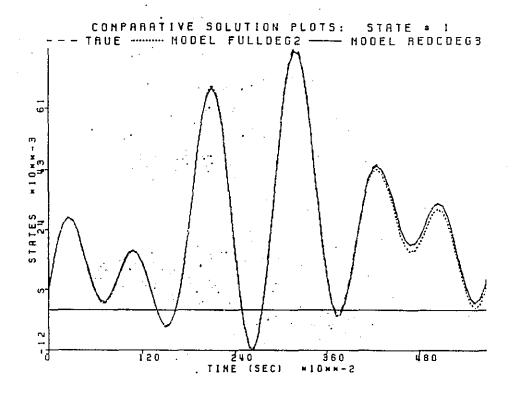


Figure 5.1.22 Simulation Number 5 of Table 5.1.14



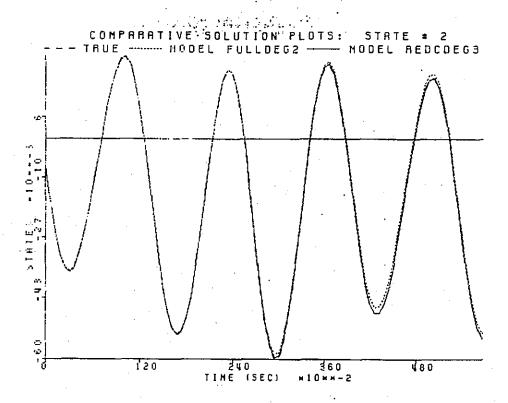


Figure 5.1.23 Simulation Number 33 of Table 5.1.14

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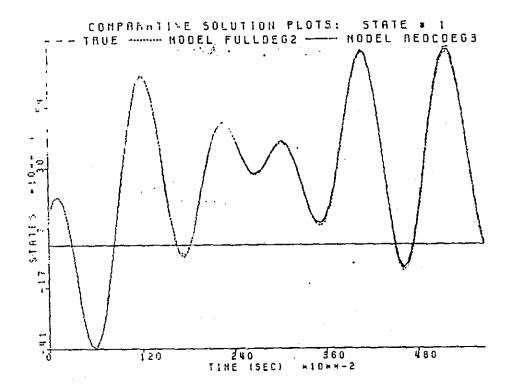
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S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1 0.005 0.005 0.005 0.005 3 0.005 0.005 4 0.005 0.005 6 0.005 0.005 6 0.005 0.005 9 0.005 0.005 10 0.005 -0.005 11 0.005 -0.005 12 0.005 -0.005 13 0.005 -0.005 14 0.005 -0.005 15 0.005 -0.005 16 0.005 -0.005 17 0.005 -0.005 18 0.005 -0.005 19 -0.005 -0.005 20 -0.005 -0.005 21 -0.005 -0.005 22 -0.005 -0.005 22 -0.005 -0.005 23 -0.005 -0.005 24 -0.005 -0.005 25 -0.005 -0.005 27 -0.005 0.005 28 -0.005 0.005 29 -0.005 0.005 31 -0.005 0.005 32 -0.005 0.005 33 -0.005 0.005 34 -0.005 0.005 35 -0.005 0.005	0.000 0.050 0.050 -0.050 -0.050 -0.050 -0.050 -0.150 0.150 -0.150 0.150 -0.150 -0.150 0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.150 0.150 0.150 0.150 0.150 0.150 -0.150 -0.150 -0.150 0.150 -0.150 -0.150 -0.150	0.75 1.00 0.75 0.75 1.00 0.00 0.00 0.00 0.00 0.00 0.00 0.0	0.101E-01 0.899E-02 0.164 0.345 0.966E-01 0.226 0.344 0.681 0.155 0.954E-01 0.260 0.133 0.153 0.929E-01 0.227 0.118 0.206E-01 0.148E-01 0.564 0.980 0.515 1.28 0.136 0.898E-01 0.158 0.898E-01 0.158 0.898E-01 0.158 0.898E-01 0.158 0.966E-01 0.158 0.966E-01 0.316 0.595 0.300 0.458 0.272 0.415 0.447 1.26 0.316 0.898E-01 0.139 0.898E-01 0.158 0.966E-01 0.1595 0.966E-01 0.158 0.966E-01 0.158 0.966E-01 0.158 0.966E-01 0.1595 0.966E-01 0.1595 0.966E-01 0.1595 0.966E-01 0.1596 0.970E-01 0.147 0.970E-01 0.148 0.907E-01 0.146 0.907E-01 0.146 0.907E-01 0.146 0.907E-01

Simulation Table for Second Degree Full Model versus Third Degree Reduced Model Table 5.1.15



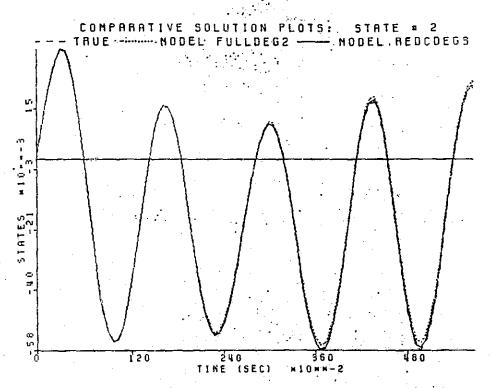


Figure 5.1.24 Simulation Number 18 of Table 5.1.15

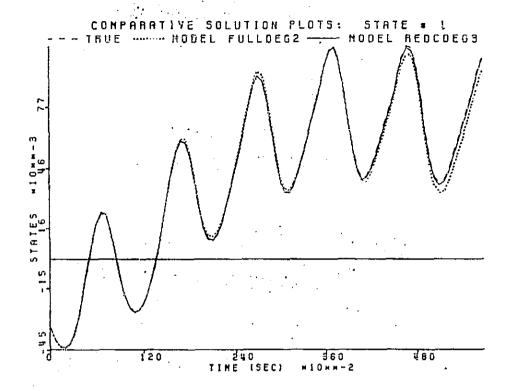
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2# INI	FIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
01234567890123456789012345678901234567	00000000000000000000000000000000000000	0.100	11111111111111111111111111111111111111	0.518E-01 0.453E-01 0.384E-01 0.251E-01 0.181E-01 0.231E-01 0.243E-01 0.250E-01 0.543E-01 0.391E-01 0.569E-01 0.391E-01 0.569E-01 0.391E-01 0.391E-01 0.341E-01 0.391E-01 0.373E-01 0.432E-01 0.373E-01 0.432E-01 0.437E-01 0.432E-01 0.437E-01 0.544E-02 0.644E-02 0.217E-01 0.169E-01 0.256E-01 0.169E-01 0.476E-01 0.255E-01 0.476E-01 0.255E-01 0.476E-01 0.178E-01 0.326E-01 0.166E-01 0.179 0.569E-01 0.179 0.559E-01 0.179 0.559E-01 0.179 0.559E-01 0.389E-01 0.166E-01 0.363E-01 0.178E-01 0.363E-01 0.178E-01 0.550E-01 0.377E-01 0.363E-01 0.195E-01 0.5552E-01 0.147E-01 0.552E-01 0.243E-01 0.5552E-01 0.243E-01 0.606E-01 0.231E-01 0.606E-01 0.231E-01 0.606E-01 0.231E-01 0.467E-01 0.231E-01 0.478E-01 0.221E-01 0.478E-01 0.221E-01 0.478E-01 0.221E-01 0.478E-01 0.221E-01 0.478E-01 0.221E-01

Table 5.1.16 Simulation Table for Second Degree Full Model versus Third Degree Reduced Model



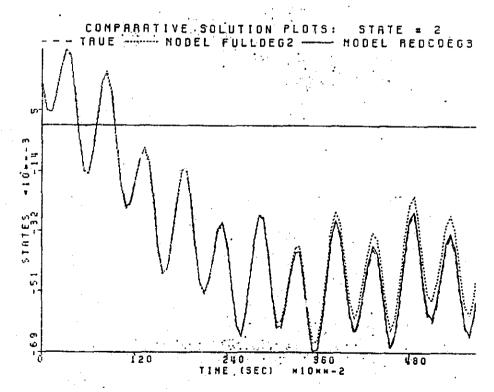
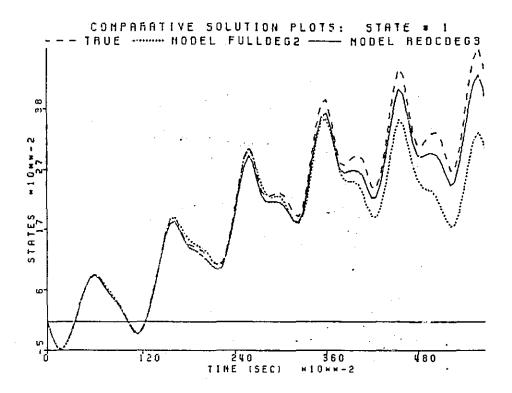


Figure 5.1.25 Simulation Number 26 of Table 5.1.16

1 0.025 0.025 0.400 0.400 2.00 1.00 2 0.025 0.025 0.400 -0.400 2.00 1.00	0,518E-01 0.282E 0.566E-01 0.295E	
2	0.103 0.485E-01 0.485E-01 0.497E-01 0.578E 0.922E-01 0.290E 0.922E-01 0.290E 0.501E-01 0.290E 0.902E-01 0.503E 0.957E-01 0.439E-01 0.300E 0.442E-01 0.442E-01 0.450E 0.268E-01 0.233E-01 0.231E-01 0.399E 0.399E 0.333E-01 0.175E 0.483E-01 0.231E-01 0.231E-01 0.399E	-011 -011 -0

Table 5.1.17 Simulation Table for Second Degree Full Model versus Third Degree Reduced Model



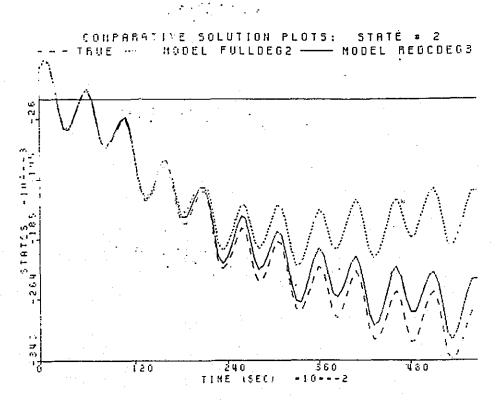
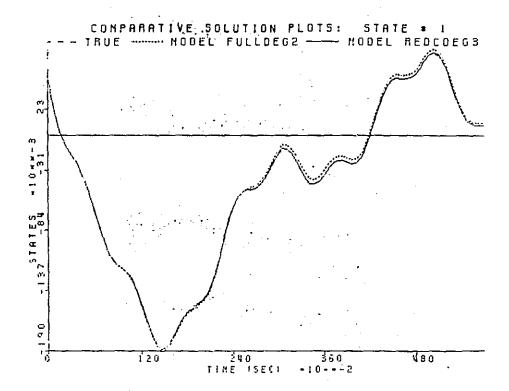


Figure 5.1.26 Simulation Number 4 of Table 5.1.17

1 0.075 0.075 9.100 0.100 0.50 1.50	0.473E-02 0.661E-02 0.444E-02 0.617E-02
1 0.075 0.075 0.100 0.100 0.50 1.50 3 0.075 0.075 0.100 0.100 0.50 1.50 4 0.075 0.075 0.100 0.100 0.50 1.50 5 0.075 0.075 0.100 0.100 0.50 1.50 6 0.075 0.075 0.100 0.100 0.50 1.50 7 0.075 0.075 0.100 0.100 0.50 1.50 8 0.075 0.075 0.100 0.100 0.50 1.50 9 0.075 0.075 0.100 0.100 0.50 1.50 10 0.075 0.075 0.100 0.100 0.50 1.50 11 0.075 0.075 0.100 0.100 0.50 1.50 12 0.075 0.075 0.100 0.100 0.50 1.50 13 0.075 0.075 0.100 0.100 0.50 1.50 14 0.075 0.075 0.100 0.100 0.50 1.50 15 0.075 0.075 0.100 0.100 0.50 1.50 16 0.075 0.075 0.100 0.100 0.50 1.50 17 0.075 0.075 0.100 0.100 0.50 1.50 18 0.075 0.075 0.100 0.100 0.50 1.50 19 0.075 0.075 0.200 0.200 0.50 1.50 19 0.075 0.075 0.200 0.200 0.50 1.50 19 0.075 0.075 0.200 0.200 0.50 1.50 21 0.075 0.075 0.200 0.200 0.50 1.50 22 0.075 0.075 0.200 0.200 0.50 1.50 23 0.075 0.075 0.200 0.200 0.50 1.50 24 0.075 0.075 0.200 0.200 0.50 1.50 25 0.075 0.075 0.200 0.200 0.50 1.50 26 0.075 0.075 0.200 0.200 0.50 1.50 27 0.075 0.075 0.200 0.200 0.50 1.50 28 0.075 0.075 0.200 0.200 0.50 1.50 29 0.075 0.075 0.200 0.200 0.50 1.50 21 0.075 0.075 0.200 0.200 0.50 1.50 22 0.075 0.075 0.200 0.200 0.50 1.50 23 0.075 0.075 0.200 0.200 0.50 1.50 24 0.075 0.075 0.200 0.200 0.50 1.50 25 0.075 0.075 0.200 0.200 0.50 1.50 26 0.075 0.075 0.200 0.200 0.50 1.50 27 0.075 0.075 0.200 0.200 0.50 1.50 28 0.075 0.075 0.200 0.200 0.50 1.50 29 0.075 0.075 0.200 0.200 0.50 1.50 30 0.075 0.075 0.200 0.200 0.50 1.50 31 0.075 0.075 0.200 0.200 0.50 1.50 31 0.075 0.075 0.200 0.200 0.50 1.50 32 0.075 0.075 0.200 0.200 0.50 1.50 33 0.075 0.075 0.200 0.200 0.50 1.50 34 0.075 0.075 0.200 0.200 0.50 1.50 35 0.075 0.075 0.200 0.200 0.50 1.50 36 0.075 0.075 0.200 0.200 0.50 1.50 37 0.075 0.075 0.200 0.200 0.50 1.50 38 0.075 0.075 0.200 0.200 0.50 1.50 39 0.075 0.075 0.200 0.200 0.50 1.50 30 0.075 0.075 0.200 0.200 0.50 1.50 31 0.075 0.075 0.200 0.200 0.50 1.50 31 0.075 0.075 0.200 0.200 0.50 1.50 31 0.075 0.075 0.200 0.200 0.50 1.50 31 0.075 0.075 0.200 0.200 0.50 1.50 31 0.075 0.075 0.200 0.200 0.50 1.50 31 0.075 0.075 0.20	0.103 0.994E-01 0.994E-01 0.665E-02 0.695E-02 0.675E-02 0.172E-02 0.165E-02 0.173E-02 0.1231E-01 0.286E-01 0.425E-01 0.405E-02 0.103E-01 0.405E-01 0.154E-01 0.154E-01 0.154E-01 0.154E-01 0.154E-01 0.154E-01 0.237E-01 0.406E-01 0.575E-01

Table 5.1.18 Simulation Table for Second Degree Full Model versus Third Degree Reduced Model



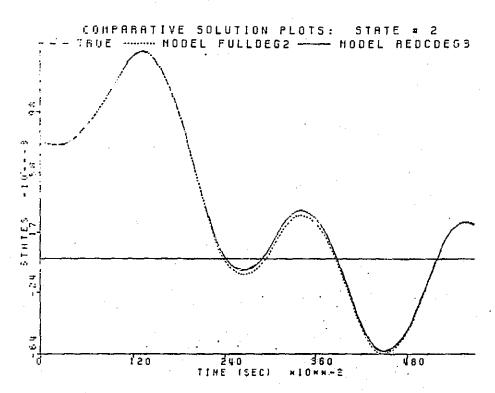
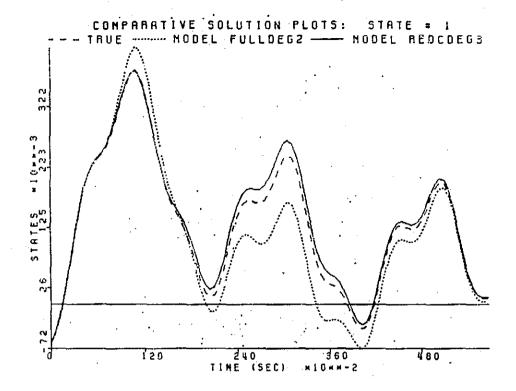


Figure 5.1.27 Simulation Number 2 of Table 5.1.18

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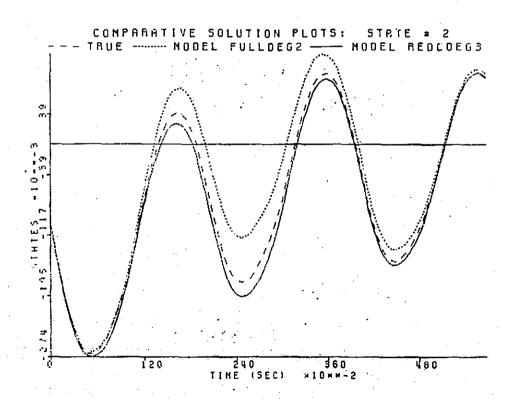
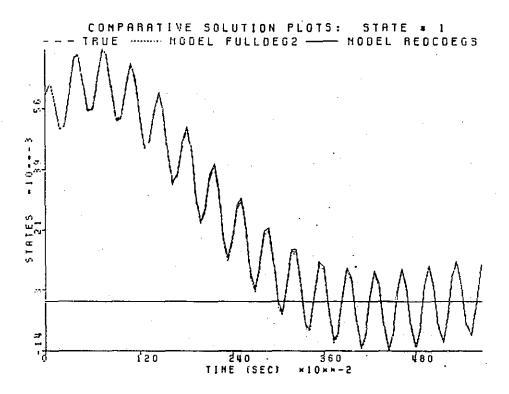


Figure 5.1.28 Simulation Number 46 of Table 5.1.18

5\$	INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1223456780-0123456787-0123456787-012355555544444444445	0.050 -0.050 0.050 -0.050 0.050 -0.050 0.050 -0.050 0.050 -0.050	-0.0173 -0.0207 -0.0217444558 -0.02174444588 -0.021740 -0.021	53187240601398113539744170807146471189502266706155809719033354788899244053424234543131378853166791113682532423454545454525325332533253325332533	0.127 0.661E-02 0.643E-02 0.705E-01 0.309E-01 0.863E-01 0.576E-01 0.775E-01 0.125E-01 0.777E-01 0.125E-01 0.289 0.445E-01 0.271E-02 0.410E-01 0.2779-01 0.467E-01 0.2779-01 0.467E-01 0.2779-01 0.467E-01 0.463E-01 0.467E-01 0.778E-01 0.467E-01 0.778E-01 0.467E-01 0.778F-01 0.467E-01 0.778F-01 0.467E-01 0.778F-01 0.467E-01 0.247E-01 0.395E-01 0.247E-01 0.395E-01 0.247E-01 0.467E-01 0.257E-01 0.284E-01 0.277E-01 0.284E-01 0.277E-01 0.467E-01 0.277E-01 0.467E-01 0.648E-01 0.288E-01 0.775E-01 0.476E-01 0.7750E-01 0.476E-01 0.4750E-01 0.476E-01 0.4750E-01 0.476E-01 0.475E-01 0.376E-01 0.4750E-01 0.376E-01 0.4750E-01 0.326E-01 0.475E-01 0.326E-01 0.4750E-01 0.326E-01 0.4750E-01 0.326E-01

Table 5.1.19 Simulation Table for Second Degree Full Model versus Third Degree Reduced Model



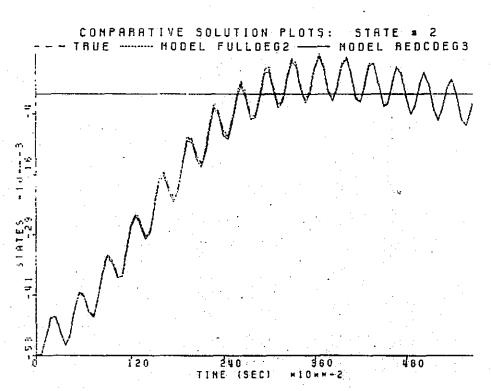
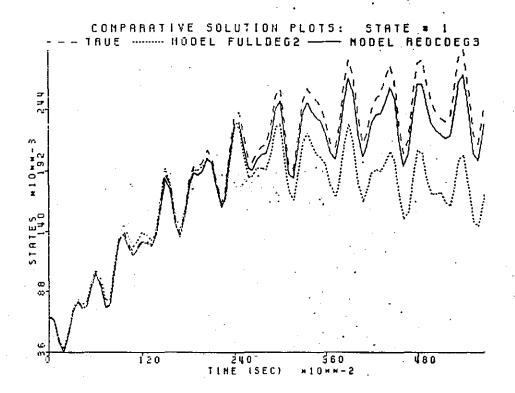


Figure 5.1.29 Simulation Number 12 of Table 5.1.19

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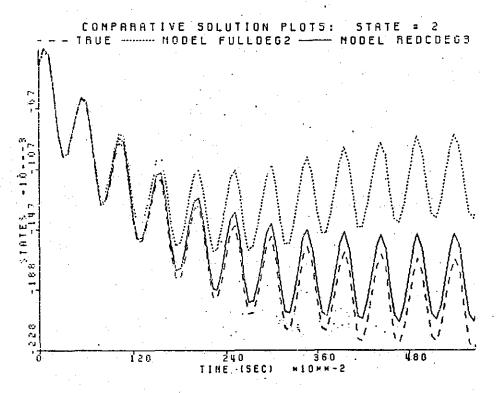
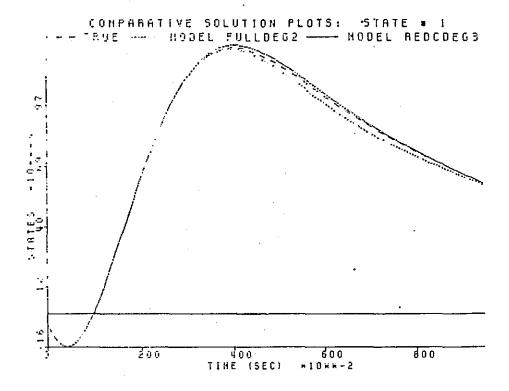


Figure 5.1.30 Simulation Number 46 of Table 5.1.19

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15

Table 5.1.20 Low Frequency Table for Second Degree Full Model versus Third Degree Reduced Model



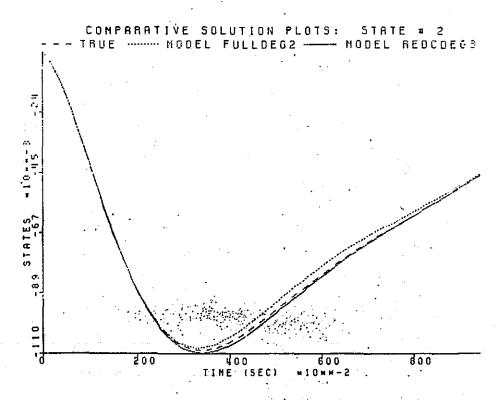
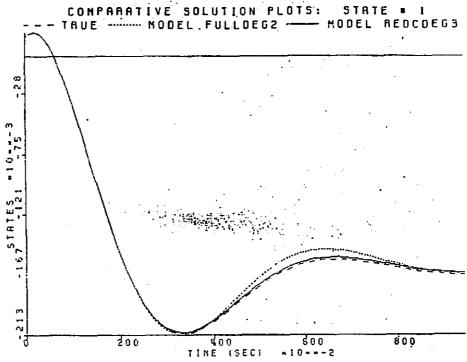


Figure 5.1.31 Simulation Number 3 of Table 5.1.20

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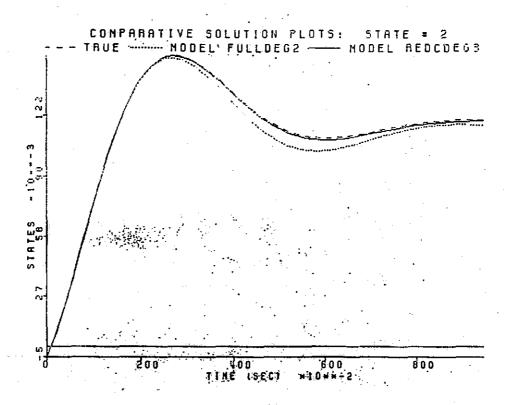
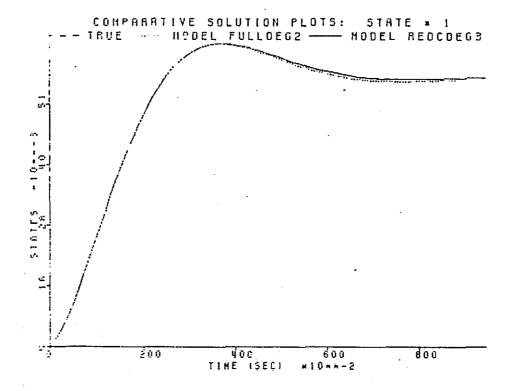


Figure 5.1.32 Simulation Number 21 of Table 5.1.20

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1	-0.022 -0.008 -0.010 -0.017 -0.022 -0.010 -0.018 -0.005 -0.021 -0.005 -0.021 -0.018 -0.021 -0.018 -0.021 -0.019 -0.017 -0.019 -0.017 -0.019 -0.015 -0.006 -0.016 -0.007 -0.016 -0.017 -0.017 -0.015 -0.017 -0.018 -0.017 -0.018 -0.017 -0.018 -0.017 -0.019 -0.019 -0.019 -0.015 -0.006 -0.015 -0.006 -0.010 -0.012 -0.010 -0.012 -0.011 -0.012 -0.011 -0.012 -0.012 -0.012 -0.012 -0.014 -0.012 -0.014 -0.015 -0.014 -0.015 -0.015 -0.015 -0.015 -0.015 -0.015 -0.015 -0.016 -0.017 -0.008	0.000 0.000	1.96 4.47 0.340E-01 0.432 0.792 0.117E-01 0.465E-02 0.237E-02 0.146E-02 35.1 0.189 5.37 4.12 1.32 0.391E-02 0.145E-02 0.2789 0.575E-01 0.255 0.575E-01 0.258 0.1462 0.887 0.575E-01 0.2580 0.142 1.59 0.575E-01 0.2589 4.64 2.39 0.395E-01 2.89 4.64 2.39 0.34.2 1.59 0.373 2.89 4.64 2.39 0.415 0.415 0.415 0.415 0.743 0.445 0.592 0.3743 0.445 0.592 0.362 0.148E-01 6.64 3.74 13.7 0.469 0.176E-01 0.819E-02

Table 5.1.21 Step Response Table for Second Degree Full Model versus Third Degree Reduced Model



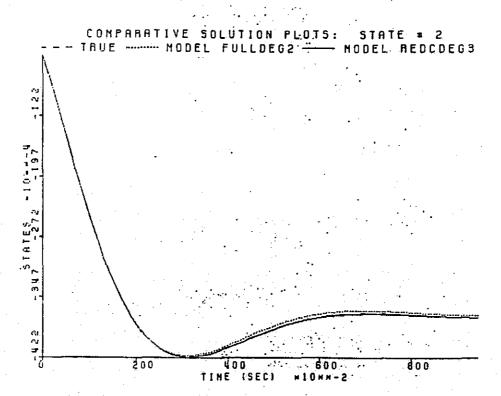
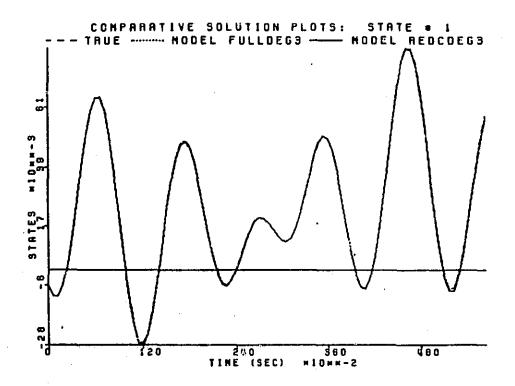


Figure 5.1.33 Simulation Number 15 of Table 5.1.21

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1 0.001 0.001 3 0.001 0.001 4 0.001 0.001 5 0.001 0.001 6 0.001 0.001 7 0.001 0.001 7 0.001 0.001 10 0.001 0.001 11 0.001 -0.001 12 0.001 -0.001 13 0.001 -0.001 14 0.001 -0.001 15 0.001 -0.001 15 0.001 -0.001 16 0.001 -0.001 17 0.001 -0.001 18 0.001 -0.001 19 -0.001 -0.001 19 -0.001 -0.001 20 -0.001 -0.001 21 -0.001 -0.001 22 -0.001 -0.001 22 -0.001 -0.001 23 -0.001 -0.001 25 -0.001 -0.001 27 -0.001 -0.001 28 -0.001 -0.001 29 -0.001 -0.001 31 -0.001 0.001 32 -0.001 0.001 33 -0.001 0.001 34 -0.001 0.001 35 -0.001 0.001	0.000 0.050 0.050 -0.050 0.050 -0.050 -0.050 0.150 0.150 -0.150 0.150 -0.150 0.150 0.050 -0.050 0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.150	0.75 1.00 0.75 1.00	0.240E-01 0.241E-01 0.487 0.859 0.497 2.30 0.711 2.00 11.1 27.6 18.1 64.9 16.8 41.7 16.8 16.0 0.239E-01 0.195E-01 0.758 3.70 0.805 5.26 0.782 3.35 13.4 16.3 17.5 48.9 17.5 48.9 17.4 0.378E-01 0.870 0.382E-01 0.870 1.14 0.903 6.05 0.726 3.4.9 1.14 4.57 0.903 6.05 0.726 34.9 1.14 4.57 0.903 6.05 0.726 34.9 1.14 4.57 0.9738 0.183E-01 0.556 2.889 0.183E-01 0.149E-01 0.5556 2.89 0.573 0.671 2.05 17.6 43.6 7.94

Table 5.1.22 Simulation Table for Third Degree Full Model versus Third Degree Reduced Model



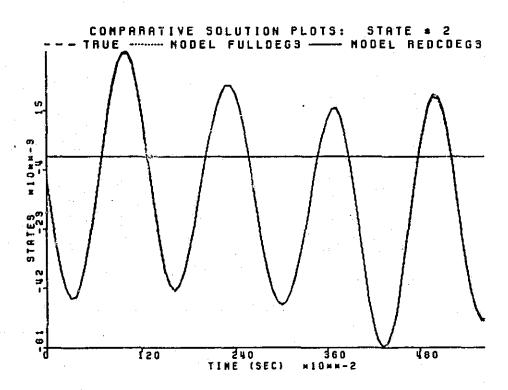
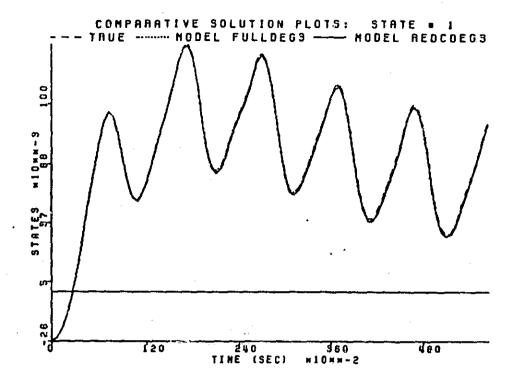


Figure 5.1.34 Simulation Number 7 of Table 5.1.22

5#	INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
123454789012345478901234547890123454789012345478901234547890123454789012345478	55555555555555555555555555555555555555	0.100	11.00000000000000000000000000000000000	0.110 0.138E-01 0.138E-01 0.1592E-0 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1592E-01 0.1692E-01 0.1792E-01 0.

Simulation Table for Third Degree Full Model versus Third Table 5.1.23 Degree Reduced Model



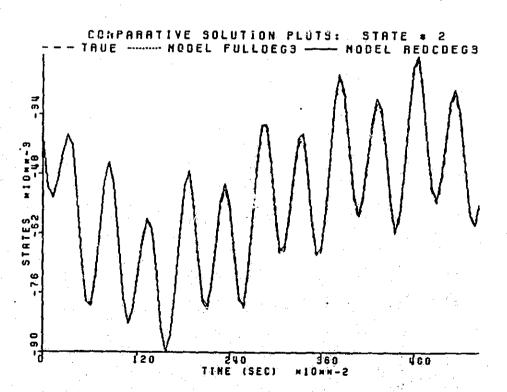
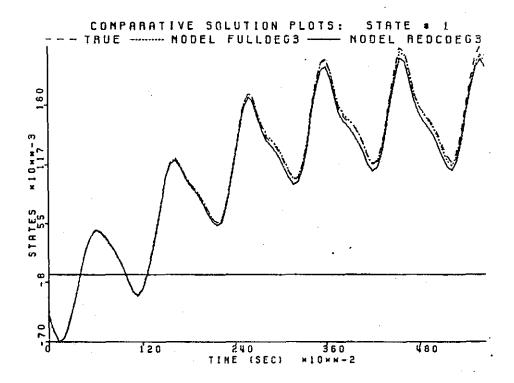


Figure 5.1.35 Simulation Number 30 of Table 5.1.23

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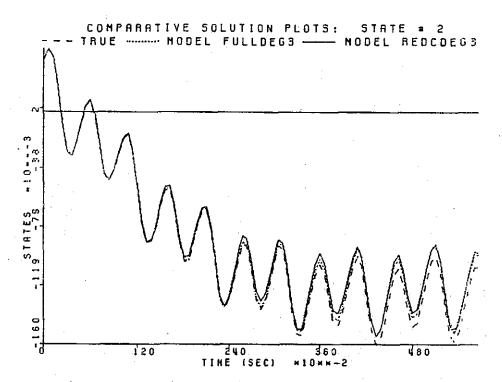
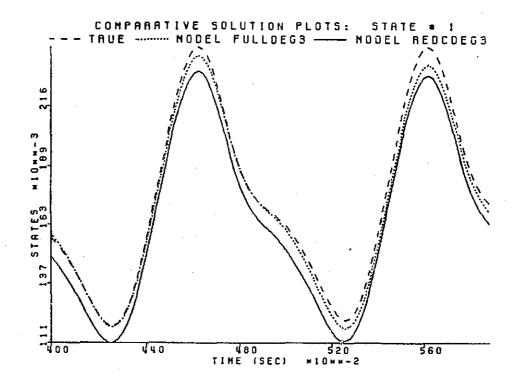


Figure 5.1.36 Simulation Number 44 of Table 5.1.23

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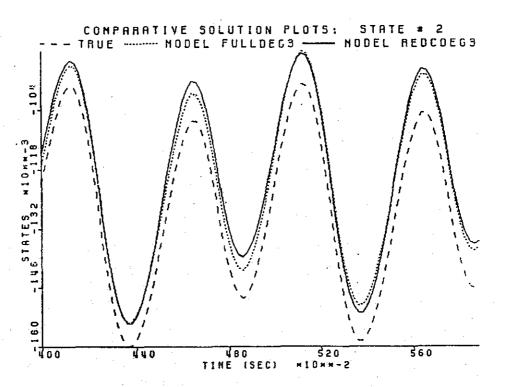
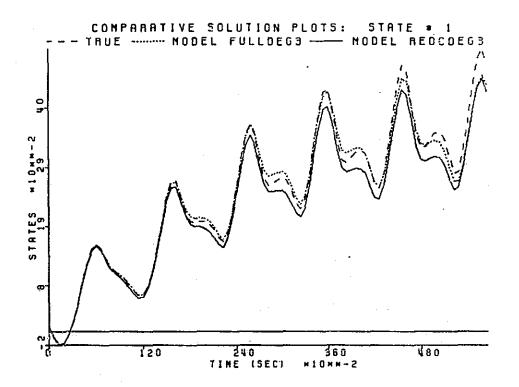


Figure 5.1.37 Expanded Section of Figure 5.1.36

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROF	RATIOS
1 0.0225555555555555555555555555555555555	0.400	11100000000000000000000000000000000000	21244776277164037891803580754788 1739805081172370011151127888875117 224412231123311112.3491511278828754788 200000000000000000000000000000000000	0.4798 4798 4798 4798 4798 4798 4798 4798

Table 5.1.24 Simulation Table for Third Degree Full Model versus Third Degree Reduced Model



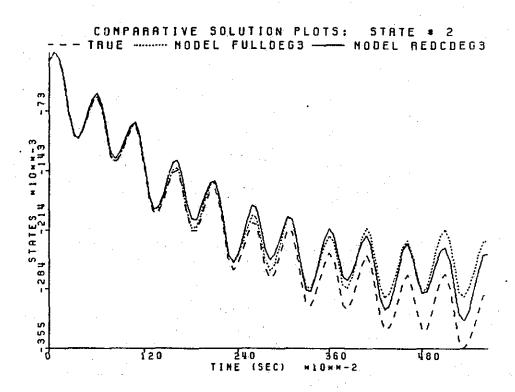
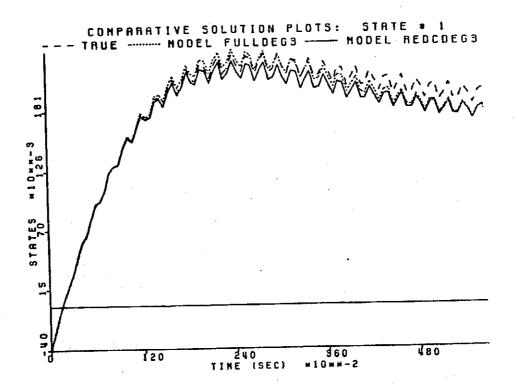


Figure 5.1.38 Simulation Number 8 of Table 5.1.24

94	INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1213456780-0123454789-01234545456789-0123454789-012345	-0.05500000555000000055500000005550000000	-0.100	64862815020542548603689730791462395283059980386 28833750206579883630435200791462395283059980386 22323342425823255444233522023442545223344432 2232334242552554442335220234425452233344432 3332619518398684647500867954060285991625006345 16110560602953158824008679540602859916210076345 44223422453445553543434252445552445535342322	1.41

Table 5.1.25 Simulation Table for Third Degree Full Model versus Third Degree Reduced Model



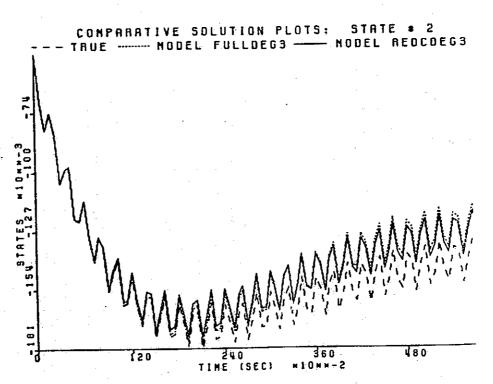
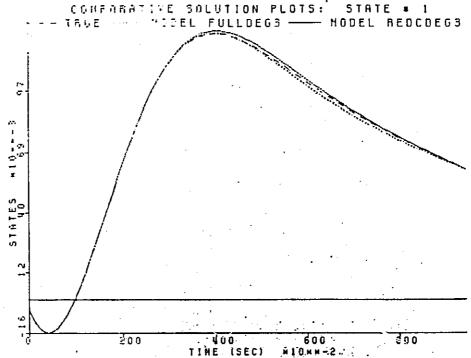


Figure 5.1.39 Simulation Number 3 of Table 5.1.25

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1 0.001 -0.001 3 0.001 -0.001 4 0.001 -0.001 5 0.001 -0.001 6 0.001 -0.001 7 0.001 -0.001 10 0.001 -0.001 11 0.001 -0.001 12 0.001 -0.001 13 0.001 -0.001 14 0.001 -0.001 15 0.001 -0.001 16 0.001 -0.001 17 0.001 -0.001 18 0.001 -0.001 19 0.010 -0.010 10 0.010 -0.010	0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.050 -0.050 0.050 -0.050 0.050 -0.050 0.010 -0.075 0.010 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.050 -0.050 0.050 -0.050 0.050 -0.050 0.050 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075	0012501250125012501250125012501250 0000000000	1.42

Table 5.1.26 Low Frequency Table for Third Degree Full Model versus Third Degree Reduced Model

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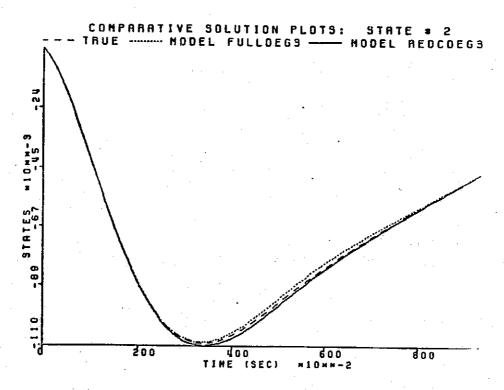
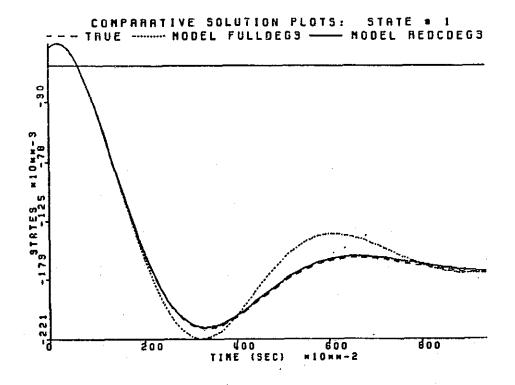


Figure 5.1.40 Simulation Number 3 of Table 5.1.26



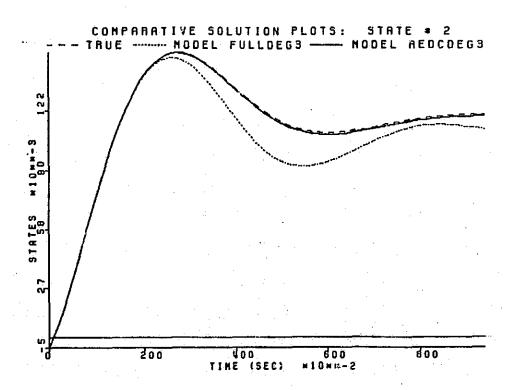
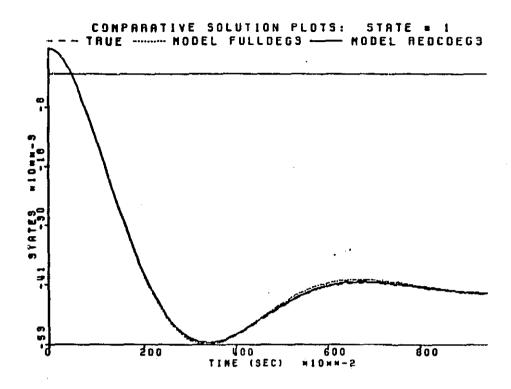


Figure 5.1.41 Simulation Number 21 of Table 5.1.26

5\$	INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
12345678901234567890123456789012345678901234567890	0.003 -0.009 -0.000 0.004 0.008 -0.005 -0.009 -0.008 0.003 -0.007 0.001 0.007 -0.008 0.004 -0.001 -0.002 -0.007 0.009 0.006 -0.005	-0.022 -0.018 -0.005 -0.012 -0.012 -0.013 -0.013 -0.013 -0.013 -0.012 -0.013 -0.012 -0.012 -0.013 -0.012 -0.013 -0.012 -0.015 -0.015 -0.015 -0.012 -0.014 -0.012 -0.015 -0.012 -0.015 -0.012 -0	0.000 0.000	0.105

Table 5.1.27 Step Response Table for Third Degree Full Model versus Third Degree Reduced Model



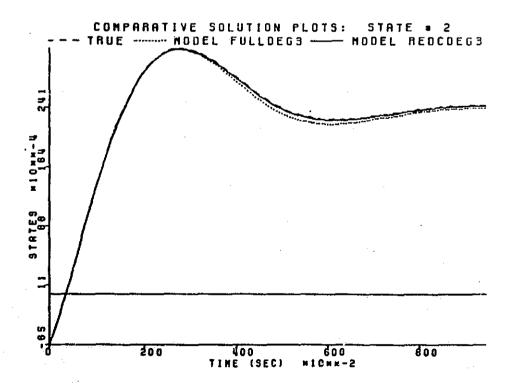


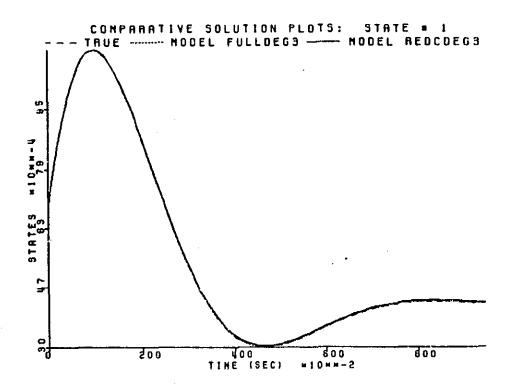
Figure 5.1.42 Simulation Number 13 of Table 5.1.27

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- Comment

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Carried Statement



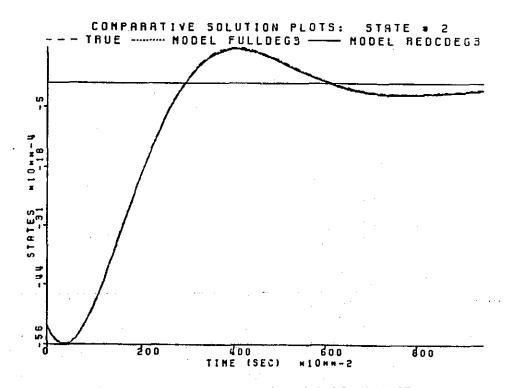


Figure 5.1.43 Simulation Number 49 of Table 5.1.27

duced model behavior is actually better than the full model behavior. So a compromise between the full and reduced models is met, the full model is stable for bigger amplitudes (that is, a larger region of stability), but the reduced model has better low frequency behavior. It is not the purpose here to resolve this decision, but rather to demonstrate that a significantly reduced model can approximate the full model without much loss of higher degree dynamics. Indeed, this was shown.

5.2 THREE STATE, THREE INPUT SYSTEM

In this example, the second degree approximation of a third degree, strictly polynomial, three state system is identified and reduced. The system is

$$\dot{x}_1 = -\frac{1}{2} x_1 + u_3 - 4x_2 - 2x_1x_3 + x_2u_1,$$

$$+ 3x_1x_2x_3 + 8x_3u_1u_2 - 5u_3,$$

$$\dot{x}_2 = -x_2 + u_1 + 3x_1x_2 - 2x_2x_3 + u_1u_2$$

$$+ 3x_1 - 2x_2x_3 + 4x_2u_2,$$

$$\dot{x}_3 = -\frac{3}{2} x_3 + u_2 - 5x_2x_3 + u_1u_3 - 2x_3u_2$$

$$+ 8x_1 - 2x_1x_2 + 3x_1u_1.$$

Again the origin is the point of expansion since it is a stable, equilibrium point. The full second degree approximation will contain terms associated with the following 27 products

x1, x2, x3, u1, u2, u3, x1x1, x1x2, x1x3, x2x2, x2x3, x3x3, x1u1, x1u2, x1u3, x2u1, x2u2, x2u3, x3u1, x3u2, x3u3, u1u1, u1u2, u1u3, u2u2, u2u3, u3u3.

A good full model was identified at the initial state conditions (0.03, 0.04, 0.04), and excited with inputs

 $u_1(t) = 0.025 \cos(2\pi t \cdot \phi_1),$ $u_2(t) = 0.015 \cos(2\pi t \cdot \phi_2),$ $u_3(t) = 0.02 \cos(2\pi t \cdot \phi_3),$

where (ϕ_1, ϕ_2, ϕ_3) were determined via optimization to be (0.7099, 1.9618, 2.9415) given the starting frequencies (0.75, 1, 3). This starting model and its corresponding reduction test appear in Figure 5.2.1. Directing our attention to this reduction test, we note that there are 11 second degree terms which appear to be good candidates for removal. Note also that the 3 linear input terms appear relatively insignificant, but it was already decided that the linear terms (both state and input terms) should be kept to preserve good local behavior. The 11 terms removed from the model structure are

x1u1, x1u2, x1u3, x2u2, x2u3, x3u1, x3u3, u1u1, u2u2, u2u3, u3u3.

Note that all terms corresponding to the squares of the states were kept.

These terms seem to reflect the appearance of the third degree system nonlinearities. This behavior was also noted in [2] where the example was first
studied. The significance of these terms is obvious.

After removal of the set of 11 terms, re-identification was performed. The same initial state conditions and input amplitudes were used, and an optimization over frequencies was performed. The starting frequencies were those used to identify the full model and the local optimization routine yielded a frequency set very near the original set. Again, this bodes well,

ENTER INITIAL CONDITIONS FOR THE 3 STATES: 0.0300 0.0400 0.0400 ENTER SAMPLING PARAMETERS: APPROXIMATION DEGREE # SAMPLE FOINTS INTERVAL BETWEEN SAMPLES INTEGRATION STEPSIZE 0.0600 0.0075 100 SHOULD THE EXCITATION BE SINUSOIDAL OR COSINUSOIDAL?[5/C]: IDENTIFICATION WILL BE DONE WITH COSINES. ENTER THE NUMBER OF (CO)SINUSCIDS PER INPUT: ENTER THE 3 FREQUENCY WEIGHTS:
0.7099 1.9618 2.9413 SHOULD THE IDENTIFICATION BE PERFORMED USING DATA THAT IS REDUCED?CY/NJ: ENTER OPTIMIZATION OPTION:

1 - NO OPTIMIZATIONS
2 - AMPLITUDES ONLY
3 - FREQUENCIES ONLY - FREQUENCIES UNLT
- INITIAL CONDITIONS DNLY
- AMPLITUDES AND FREQUENCIES
- AMPLITUDES AND INITIAL CONDITIONS
- FREQUENCIES AND INITIAL CONDITIONS
- AMPLITUDES, FREQUENCIES, AND INITIAL CONDITIONS DO YOU WISH TO NORMALIZE THE DATA?[Y/N]: THE MATRIX OF SAMPLED MONOMIAL TERMS HAS 27 ROWS AND 100 COLUMNS.

NUMBER OF TIMES COST FUNCTION WAS EVALUATED: WOULD YOU LIKE TO SEE PLOTS OF THE INPUTS? IF SO, LINE UP THE CARRIAGE.

TSO SPEAKEASY III FI+ 11:13 FM FEBRUARY 29, 1984 :_SIZE=500;GET IDENT;IDENT;QUIT

EXECUTION STARTED

Figure 5.2.1a Full Second Degree Model and Reduction Test (First Pass)

```
FARTITION NUMBER 1.
FRITITION (A 3 BY 3 ARRAY)
-.50043 -2.9762E-4 -.0028604
-6.0605E-4 -.99939 4.3545E-4
-6.5976E-5 -5.3277E-4 -1.4982
  ...WITH EIGENVALUES:
VALUES (A VECTOR WITH 3 COMPONENTS)
-1.4982 -.99939 -.50043
 PARTITION NUMBER 2.
PRTITION (A 3 BY 3 ARRAY)
9.1863E-5 -6.8572E-4 .9969
.99983 -3.1204E-5 -1.928E-5
4.0466E-6 .99868 -1.1076E-6
 PARTITION NUMBER 3.
PRITITION (A 3 BY 6 ARRAY)
.053613 .037125 -1.7381 -3.9029 -.49871
.084004 3.0326 -.15235 -.042012 -1.9158
.097414 .23472 -.21056 -.10562 -5.2205
                                                                                                                                 •3702?
•054736 •
                                                                                                                                 37405
PARTITION NUMBER 4.
PRITITION (A 3 BY 9 ARRAY)
ROW 1 -.028705 -.042196 -.040007 1.0099 .022295
.0030023 .071073
ROW 2 -.0035305 .012927 .022232 .0044049 -.0032042
-.014618 -.058518
ROW 3 .0079307 -.033548 -.0011774 .0056223 .034948
-2.0196 -.0037745
                                                                                                                                                               .011026
                                                                                                                                                                                          .017238
                                                                                                        .0044049 -.0032042 -.018316 -.010389
                                                                                                                                                              .0045049 -.049517
PARTITION NUMBER 5.
PRITITION (A 3 BY 6 ARRAY)
.0071867 .01124 -4.6141E-5 .0028277
-.0026629 .99769 .001181 .020343
.01167 -.0027814 .9962 -.0015025
                                                                                                                      -.0037576 -.01968
-2.9842E-4 -.0010876
-.0041424 -.0015666
S (A 27 COMPONENT ARRAY)
10.112 7.5281 7.4573 7.3082
3.5474 3.5094 2.6157 2.5262
.51718 .48339 .43195 .39965
.044619 .025358 .029926
THE MAXIMUM SINGULAR VALUE:
MAX = 10.112
THE MINIMUM NONZERO SINGULAR VALUE:
MIN = .025758
                                                                                                       5.5485
2.4238
.31808
                                                                                                                               5.0816
1.1098
.30539
                                                                                                                                                        4.972
.82666
.17233
 MIN = .025358
...AND THEIR RATIO: RTIO = 398.76
 WANT TO TRY REDUCTION TEST?[Y/N]: Y
FULL SYSTEM ERROR:
FULL STATE ERRORS:
                                                                                                  40.8687
TERM: X1. COLUCHANGE IN SYSTEM ERROR: CHANGE IN STATE ERRORS:
                                                 COLUMN #: 1.
RROR: -29.4887
                                                                         -29,4959
                                                                                                                  0.0107
                                                                                                                                                  -0.0035
 TERM:
                                     X2. COLUMN #:
                                                                                    2.
```

Figure 5.2.1b Full Second Degree Model and Reduction Test (First Pass)

CHANGE CHANGE	IN IN	SYSTEM ERROR: STATE ERRORS:	26,2132 -0,0187	26.2604	-0.0286
TERM: CHANGE CHANGE	IN IN	X3. COLUMN SYSTEM ERROR: STATE ERRORS:	7: 3. -55.0800 -0.1408	-0.0054	-54.9338
TERM: CHANGE CHANGE	IN IN	U1. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 4. 6.0935 0.0001	6.0933	0.0001
		U2. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 5, -3.9866 0.0002	-0.0001	-3,9866
TERM: CHANGE CHANGE	IN IN	U3. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 6. 0.9497 0.9499	-0.0001	-0.0000
TERM: CHANGE CHANGE	IN IN	X1.X1. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 7. 6.0076 2.7131	-1.0940	4.3885
TERM: CHANGE CHANGE	IN NI	X1.X2. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 8. 18.7248 1.8695	6.0373	10.8180
TERM: CHANGE CHANGE	IN IN	X1.X3. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 9. -66.5199 -60.3416	1,6895	-7,8678
TERM: CHANGE CHANGE	NI NI	X2.X2. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 10. -125.1554 -120.9146	0.5453	-4.7861
TERM: CHANGE CHANGE	IN IN	X2.X3. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 11. -93.8660 -19.7353	33.1097	-107,2404
TERM: CHANGE CHANGE	IN IN	X3.X3. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 12. 25.6494 13.1292	-0.4948	13.0150
TERM: CHANGE CHANGE	IN IN	X1.U1. COLUMN SYSTEM ERROR: STATE ERRORS:	-0.0621	0.0428	0.1405
TERM: CHANGE CHANGE	IN IN	X1.U2. COLUMN SYSTEM ERROR: STATE ERRORS:	14. 0.2876 0.0456	0.0120	0.2300
TERM: CHANGE CHANGE	IN IN	X1.U3. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 15. 0.5834 0.5657	0.0146	0.0031
TERM: CHANGE CHANGE	IN IN	X2.U1. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 16. 16.1320 16.0601	-0.0387	0.1105
TERM: CHANGE CHANGE	IN IN	X2.U2, COLUMN SYSTEM ERROR: STATE ERRORS:	#: 17. -0.1156 -0.0386	0+0019	-0.0788
TERM: CHANGE CHANGE	IN IN	X2.U3. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 18. -0.1334 -0.1107	-0.0002	-0.0225
TERM: CHANGE		X3.U1. COLUMN SYSTEM ERROR:	#; 19. -0.6325		

Figure 5.2.lc Full Second Degree Model and Reduction Test (First Pass)

CHANGE	IN	STATE ERRORS:	0.2598	0.0735	-0.9658
TERM: CHANGE CHANGE	IN IN	X3.U2. COLUMN SYSTEM ERROR: STATE ERRORS:	1: 20. 8.5980 -0.0023	0.0002	8.6001
		X3.U3. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 21. -0.4704 -0.5187	0.0362	0.0121
TERM: CHANGE CHANGE	IN IN	U1.U1. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 22. 0.7705 0.3057	0.0370	0.4278
TERM: CHANGE CHANGE	IŅ IN	U1.U2. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 23, -22,4779 -0,0165	-22.4503	-0.0111
TERM: CHANGE CHANGE	IN IN	U1.U3. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 24. -21.4666 0.0001	0.0028	-21,4695
TERM: CHANGE CHANGE	IN IN	U2.U2. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 25. -0.1601 0.1269	-0.2379	· -0.0492
TERM: CHANGE CHANGE	IN IN	U2.U3. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 26. 0.0087 0.0146	-0.0017	-0.0042
		U3.U3. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 27. -0.9049 -0.8697	0.0129	-0.0481

DO YOU WANT TO DISCARD ANY TERMS AND RE-OFTIMIZE?[Y/N]: Y

HOW MANY TERMS WILL BE KEPT? 16

ENTER THE COLUMN NUMBERS OF COLUMNS WHICH ARE TO BE KEPT.
ENTRY OF COLUMN NUMBERS WILL TAKE PLACE IN BLOCKS OF 10.
WHEN ASKED, ENTER A AS AN ARRAY WITH A MAXIMUM OF TEN ELEMENTS.
A = ARRAY(10:1,2,3,4,5,6,7,8,9,10)

A = ARRAY(6:11,12,16,20,23,24)

MANUAL MODE SPACE USED 99 K NOW, 118 K PEAK, SIZE 500 K
DO YOU WISH TO SAVE THIS MODEL?[Y/N]:N
DO YOU WISH TO IDENTIFY ANOTHER MODEL?[Y/N]:N

Figure 5.2.1d Full Second Degree Model and Reduction Test (First Pass)

since we do not wish much change. The resulting sixteen term model and its reduction test appear in Figure 5.2.2. With the exception of the linear control inputs, there are only two terms which appear to be the next candidates for removal. They are the x₁x₁ term and the x₃u₂ term. These terms were removed and the model was re-identified, but simulations showed too much decline in model performance, especially in the third state. (This loss of performance in the third state might have been anticipated since the error test for the third state for each of the two terms contained a more significant entry.) So, we take the 16 term model as the final one. The reduction then is 11 terms removed out of a possible 27; a 41% reduction.

Simulation will be presented in the same fashion as the presentation made for the previous example of second degree model reduction. That is, several tables of simulations of the exact linear model versus the second degree reduced model are given. Following that, simulation tables for the second degree full model versus the second degree reduced model are presented. In all cases, each table is followed by at least one set of plots picked as a representative of those simulations in the table. To facilitate the output, only two of the three states are shown. Usually this includes a worst case and a typical comparison plot.

Tables 5.2.1 through Table 5.2.5 are simulation tables of the exact linear model versus the reduced second degree model. In each of these tables MODEL1 refers to the reduced model, and the error ratio is the mean square error of the reduced model over the mean square error of the linear model. The linear model for this system is

$$\mathbf{L}_{10} = \begin{bmatrix} -0.5 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -1.5 \end{bmatrix}, \quad \mathbf{L}_{01} = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}.$$

Table 5.2.1 and Table 5.2.2 contain simulations which hold the frequencies constant while the cosinusoidal input amplitudes are varied along with the initial conditions. Table 5.2.3 studies the models at higher frequencies. Table 5.2.4 varies the control amplitudes and frequencies while holding the initial conditions constant. Note that the third state error ratio of simulation 3 indicates that the linear model outperformed the reduced model. This does not cause much concern as the plot shows the two curves are actually very close. Table 5.2.5 is a low frequency table. The corresponding plots show a typical step response and the worst case response to a very low frequency input. Despite an occasional simulation instance, in each of the five tables, the reduced second degree model outperforms the linear model satisfactorily.

As the final test, the reduced model is compared against the full second degree model from which it was obtained. Now in these tables the labels MODEL1 and MODEL2 refer to the full and reduced models respectively. The error ratios are the mean square errors of the reduced model over the mean square error of the full model. The same five tables of simulations as were calculated previously will be run here. To review, Table 5.2.6 and Table 5.2.7 vary the initial conditions and input amplitudes while the frequencies are held constant. Table 5.2.8 contains simulations of higher frequencies. Table 5.2.9 varies the frequencies while the initial state conditions are held constant. Finally Table 5.2.10 demonstrates the low frequency behavior of the models. Note that in all cases the 16 term model is a good approximation of the full 27 term model. Again it appears that the reduction method works quite well as both of the criteria for judging reduced models were met.

```
ENTER INITIAL CONDITIONS FOR THE 3 STATES:
         0.0300
                            0.0400
 ENTER SAMPLING PARAMETERS:
APPROXIMATION DEGREE
# SAMPLE POINTS
                 INTERVAL BETWEEN SAMPLES INTEGRATION STEPSIZE
         100
                      0.0600
  SHOULD THE EXCITATION BE SINUSOIDAL OR COSINUSOIDAL?[5/C]:
  IDENTIFICATION WILL BE DONE WITH COSINES.
  ENTER THE NUMBER OF (CO)SINUSOIDS PER INPUT:
   ENTER THE 3 FREQUENCY WEIGHTS: 0.7099 1.9618 2
         0.7099
  SHOULD THE IDENTIFICATION BE PERFORMED USING DATA THAT IS REDUCED?[Y/N]:
 ENTER_OPTIMIZATION OPTION:
1 - NO OPTIMIZATIONS
2 - AMPLITUDES_ONLY
               2 - AMPLITUDES ONLY
3 - FREQUENCIES ONLY
4 - INITIAL CONDITIONS ONLY
5 - AMPLITUDES AND FREQUENCIES
6 - AMPLITUDES AND INITIAL CONDITIONS
7 - FREQUENCIES AND INITIAL CONDITIONS
8 - AMPLITUDES, FREQUENCIES, AND INITIAL CONDITIONS
3
 ENTER THE TRACE OF THE COVARIANCE MATRIX: 0.9999999999999954E-06
 ENTER THE OPTIMIZATION METHOD: 1,2,3 OR 4
 ENTER # OF SIGNIFICANT DIGITS FOR CONVERGENCE, MAXIMUM # OF FUNCTION CALLS, AND THE ZXMIN OPTION # (0,1,2,3) 2 200 2
- 20 CALLS OF FCN....
THE PARAMETERS:
0.7076123D+00 0.1970220D+01 0.2940915D+01
THE CONDITION NUMBER: 0.313D+08
DBJECTIVE FUNCTION VALUE: 0.1669D+04
- 40 CALLS OF FCN....
 0.70873390400 0.19696630401 0.29417480401
THE CONDITION NUMBER: 0.3130408
OBJECTIVE FUNCTION VALUE: 0.16680404
   60 CALLS OF FCN....
 THE PARAMETERS: 0.7086491D+00 0.1969523D+01 0.2941611D+01
 THE CONDITION NUMBER: 0.313D+08
OBJECTIVE FUNCTION VALUE: 0.1668D+04
   Figure 5.2.2a Reduced Second Degree Model and Reduction Test
```

(Second Pass)

```
CONVERGENCE WAS ACHIEVED AND NO ERRORS OCCURRED.
```

DO YOU WISH TO NORMALIZE THE DATA? CY/NJ:

THE MATRIX OF SAMPLED MONOMIAL TERMS HAS 16 ROWS AND 100 COLUMNS.

NUMBER OF TIMES COST FUNCTION WAS EVALUATED:

THE OPTIMUM FREQUENCIES ARE: FREQ(1)= 0.709 FREQ(2)= 1.970 FREQ(3)= 2.942

WOULD YOU LIKE TO SEE PLOTS OF THE INPUTS? IF SO: LINE UP THE CARRIAGE.

TSO SPEAKEASY III PI+ 3:02 PM MARCH 2, 1984 !_SIZE=500;GET IDENT;IDENT;QUIT EXECUTION STARTED COLSKEPT (A 16 COMPONENT ARRAY) 1 2 3 4 5 6 7 8 9 10 11 12 16 20 23 24

68

...WITH EIGENVALUES: VALUES (A VECTOR WITH 3 COMPONENTS) -1.499 -.99922 -.50041

PARTITION NUMBER 2. PRIITION (A 3 BY 3 ARRAY) 1.0112E-5 -8.3356E-4 .99691 .99981 2.3861E-5 -1.6467E-4 1.1036E-5 .99857 8.323E-7

PARTITION NUMBER 3.
PRITITION (A 3 BY 6 ARRAY)
-9.8639E-5 .14795 -1.6646 -3.9437 -.55746 .39095
.11372 2.9703 -.27546 -.05058 -1.7861 .028917
-.18827 .42369 .074939 -.20082 -5.174 .19199

PARTITION NUMBER 4.
PRITITION (A 3 BY 9 ARRAY)

ROW 1 0 0 0 .77664 0 0 0

ROW 2 -0 0 0 -.0021199 0 0 0

-.0056813 0

ROW 3 0 0 0 -.0082638 0 0 0

-2.0038 0

PARTITION NUMBER 5.
PRITITION (A 3 BY 6 ARRAY)
0 .016329 9.8794E-4 0 0 0

Figure 5.2.2b Reduced Second Degree Model and Reduction Test (Second Pass)

```
.99372 -.0011691
3.1774E-4 .99658
 S (A 16 COMPONENT ARRAY)
7.3001 6.6372 7.0392 7.1091
1.3379 .80738 .3557 .24935
THE MAXIMUM SINGULAR VALUE:
MAX = 7.3001
THE MINIMUM NONZERO SINGULAR VALUE:
MIN = .028725
                                                                              4.9926
                                                                                              4.8108
                                                                                                                  1.9791
                                                                                                                                    1.3043
                                                                                                 .052128 .028725 .036267
                                                                              .18817
 ...AND THEIR RATIO: RTIO = 254.14
 WANT TO TRY REDUCTION TEST?[Y/N]; Y
 FULL SYSTEM ERROR: FULL STATE ERRORS:
                                          326.9198
137.6369
                                                                          39.8868 ...
                                                                                                  149,3960
 TERM: X1. COLUMN #: 1.
CHANGE IN SYSTEM ERROR: -29.0252
CHANGE IN STATE ERRORS: -29.1570
                                                                                      0.0061
                                                                                                                 0.1258
 TERM: X2. COLUMN #: 2.
CHANGE IN SYSTEM ERROR: 25.4363
CHANGE IN STATE ERRORS: -0.0310
                                                                                    25.5185
                                                                                                               -0.0513
 TERM: X3. COLUMN #: 3.
CHANGE IN SYSTEM ERROR: -55.9796
CHANGE IN STATE ERRORS: -0.1721
                                                                                                             -55.7944
                                                                                    -0.0130
 TERM: U1. COLUMN #:
CHANGE IN SYSTEM ERROR:
CHANGE IN STATE ERRORS:
                                                            6.4543
                                                                                      6.4541
                                                                                                                 0.0002
 TERM: U2. COLUMN #: 5.
CHANGE IN SYSTEM ERROR: -4.0565
CHANGE IN STATE ERRORS: 0.0010
                                                                                      0.0001
                                                                                                               -4.0576
TERM: U3, COLUMN #: 6.
CHANGE IN SYSTEM ERROR: 0.8909
CHANGE IN STATE ERRORS: 0.8917
                                                                                    -0.0009
                                                                                                                 0.0000
 TERM: X1.X1. COLUMN 8: 7.
CHANGE IN SYSTEM ERROR: -9.7038
CHANGE IN STATE ERRORS: -0.0049
                                                                                    -1.3848
                                                                                                               -8.3140
 TERM: X1.X2. COLUMN #: 8.
CHANGE IN SYSTEM ERROR: 34.5059
CHANGE IN STATE ERRORS: 7.4622
                                                                                      6.9057
                                                                                                               20.1380
 TERM: X1.X3. COLUMN #:
CHANGE IN SYSTEM ERROR: -
CHANGE IN STATE ERRORS: -
                                                                                      3,0254
                                                                                                                 2.8783
 TERM: X2.X2. COLUMN #: 10.
CHANGE IN SYSTEM ERROR: -127.9159
CHANGE IN STATE ERRORS: -119.4493
                                                                                      0.6269
                                                                                                               -9.0936
 TERM: X2.X3. COLUMN #:
CHANGE IN SYSTEM ERROR: -10
CHANGE IN STATE ERRORS: -2
                                                     -101.3486
-21.7269
                                                                                    29.1325
                                                                                                          -108.7542
 TERM: X3.X3. COLUMN *: 12.
CHANGE IN SYSTEM ERROR: 20.1306
CHANGE IN STATE ERRURS: 13.7498
                                                                                    -0.2507
                                                                                                                 6.6315
```

Figure 5.2.2c Reduced Second Degree Model and Reduction Test (Second Pass)

TERM: CHANGE CHANGE	IN	X2.U1. COLUMN SYSTEM ERROR: STATE ERRORS:	15.4849 15.6285	0.0179	-0.1614
		X3.U2. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 20. 8.7552 0.0066	-0.0001	8.7487
	IN	U1.U2. COLUMN SYSTEM ERROR: STAT5 ERRORS:	#: 23. -22.3259 -0.0164	-22.3106	0.0011
		U1.U3. COLUMN SYSTEM ERROR: STATE ERRORS:	#1. 24; -21.2391 -0.0014	-0.0020	-21.2356

DO YOU WANT TO DISCARD ANY TERMS AND RE-OPTIHIZE?CY/NJ: N HANUAL HODE SPACE USED 71 K NOW, 83 K PEAK, SIZE 500 K

DO YOU WISH TO SAVE THIS MODEL?[Y/N]:N
DO YOU WISH TO IDENTIFY ANOTHER MODEL?[Y/N]:N

Figure 5.2.2d Reduced Second Degree Model and Reduction Test (Second Pass)

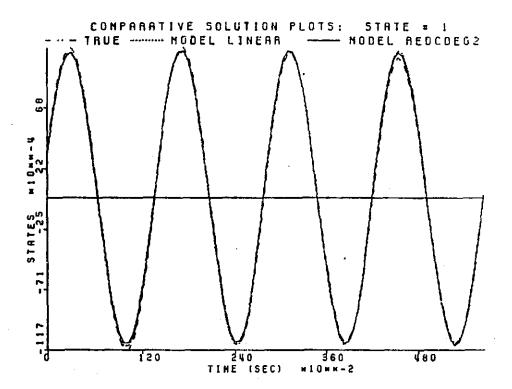
PROBLEM SUMMARY CONFIGURATION: TRUE-LINEAR-MODEL1 OF STATES: 3 OF STATES: 3 OF TEMS IN MODEL 1: 16 DEGREE OF APPROXIMATION: 2 SIMULATION WITH COSINE

St	INITIAL COMPUTI	OKS	AMPLITUDES	FREQUENCIES	ERROP RATIOS
17.345678901123456789011314	0.1009E-02 0.1000E-02 0.1009E-02 0.5009E-02	0.1009E-02 0.1000E-02 0.1000E-02 0.1000E-02 0.1000E-02 0.1000E-02 0.1000E-02 1000E-02 1000E-02 1000E-02 1000E-02 1000E-02 1000E-02 0.5000E-02 0.5000E-02 0.5000E-02 0.5000E-02 0.5000E-02 0.5000E-02 0.5000E-02	0.010 -0.010 -0.010 -0.010 -0.010 0.010 9.050 0.050 -0.050 -0.029 0.020 -0.050 -0.025 -0.075 -0.075 0.025 -0.075 0.025 0.025 -0.075 0.025 0.010 -0.010 -0.010 -0.010 -0.010 -0.010 -0.010 -0.010 0.010 0.050 0.050 -0.050 -0.050 0.050 -0.050 -0.055 0.050 -0.050 -0.055 0.055 -0.075 0.025 0.055 0.050 -0.055 0.055 0.050 -0.055 0.055 0.050 -0.055 0.055 0.050 -0.055 0.055 0.055 0.010 -0.010 0.010 0.050 0.055 0.055 0.055 0.055 0.055 0.055 0.055 0.055 0.055 0.055 0.055 0.055 0.055 0.055 0.055 0.055 0.055	0.50 0.60 0.70 0.50 0.60 0.70	0.4513E-01 0.2112E-02 0.1085E-01 0.1161 0.1327E-02 0.2264E-01 0.4214E-01 0.9359E-03 0.1961E-02 0.345EE-01 0.5510E-03 0.2247E-02 0.6242E-01 0.2275E-02 0.1788E-02 0.7869 0.27729E-02 0.2626E-02 0.1895E-02 0.283EE-02 0.2626E-02 0.1195E-01 0.1012E-02 0.6083E-02 0.6107E-01 0.575E-02 0.100E-01 0.1263 0.9690E-03 0.2176E-01 0.4358E-01 0.9690E-03 0.2176E-01 0.4358E-01 0.9399E-03 0.2176E-01 0.3449E-01 0.442E-03 0.255E-02 0.526E-01 0.224E-02 0.1803E-02 0.7819 0.2714E-02 0.2628E-02 0.1228E-01 0.1007E-02 0.6384E-02 0.5158E-01 0.1110E-01 0.6563E-01 0.1554 0.3678E-03 0.7552E-01 0.3714E-01 0.8000E-03 0.2386E-02 0.3540E-01 0.8000E-03 0.278E-02 0.3540E-01 0.8000E-03 0.278E-02 0.3540E-01 0.8000E-03 0.278EE-02 0.3540E-01 0.8000E-03 0.278EE-02 0.3540E-01 0.8000E-03 0.278EE-02 0.6275E-01 0.2274E-02 0.4420E-02 0.7672 0.2800E-02 0.2993E-02 0.9202E-02 0.3992E-02 0.1784E-02 0.1146E-01 0.1697E-02 0.4338E-02

Table 5.2.1 Simulation Table for Linear Model versus Second Degree Reduced Model

Sŧ	INITIAL CONDITIO	ONS	AMPLITUDES	FREDUENCIES	ERROR RATIOS
1234567.89.0112345167.89.01234512222222233	- \$500E-02	0.5000E-02 0.5000E-02 0.5000E-02 0.5000E-02 0.5000E-02 0.5000E-02 0.5000E-017500E-01	0.010 -0.010 -0.010 -0.010 -0.010 -0.010 0.050 0.050 0.050 -0.050 0.050 -0.050 -0.050 0.050 -0.050 -0.055 0.055 -0.075 0.025 0.025 0.025 0.025 0.025 0.025 0.010 -0.010 -0.010 -0.010 -0.010 0.010 -0.010 -0.010 0.050 -0.050 0.050 -0.050 -0.050 0.050 -0.050 -0.050 0.050 -0.050 -0.050 0.050 -0.050 -0.050 0.050 -0.050 -0.050 0.050 0.055 -0.055 0.050 0.055 -0.055 0.050 0.055 -0.055 0.050 0.055 -0.055 0.050 0.055 -0.055 0.055 0.055 -0.055 0.055 0.055	0.50 0.60 0.70 0.50 0	0.4602E-01 0.1368E-01 0.1667E-01 0.5347E-01 0.3101E-02 0.4939E-01 0.4424E-01 0.9840E-03 0.1762E-02 0.46379E-01 0.1070E-02 0.2125E-02 0.6142E-01 0.2018E-02 0.2125E-02 0.8405 0.5664E-02 0.2898E-02 0.7917E-02 0.3623E-02 0.1757E-02 0.1243E-01 0.1111E-02 0.3028E-02 0.2301E-01 0.4174E-01 0.4320E-01 0.2175E-01 0.4343E-01 0.3713E-01 0.2175E-01 0.4343E-01 0.3713E-01 0.2175E-01 0.4343E-01 0.3713E-01 0.1458E-01 0.9202E-01 0.4772E-01 0.1458E-01 0.9202E-01 0.4772E-01 0.1458E-01 0.9202E-01 0.4772E-01 0.1458E-01 0.9202E-01 0.3336E-01 0.2274E-01 0.255E-01 0.8098E-02 0.1374E-01 0.2658E-01 0.8098E-02 0.1374E-01 0.6150E-02 0.1305 0.5672E-01 0.6098E-02 0.1305 0.5672E-01 0.6098E-02 0.1305 0.46712E-01 0.6098E-02 0.1305 0.46712E-01 0.6098E-02 0.1305 0.46712E-01 0.4430E-02 0.1597 0.5542E-02 0.1905E-01 0.6717E-01 0.1074 0.1629E-01 0.2775 0.1862E-01 0.1374E-01 0.8064E-03 0.2031E-01 0.1374E-01 0.8064E-03 0.2032E-01 0.1374E-01 0.8064E-03 0.2032E-01 0.1374E-01 0.8064E-03 0.2132E-01 0.1374E-01 0.8064E-03 0.2132E-01 0.1374E-01 0.8064E-03 0.2132E-01 0.1374E-01 0.8064E-03 0.2132E-01 0.1296E-01 0.8064E-03

Table 5.2.2 Simulation Table for Linear Model versus Second Degree Reduced Model



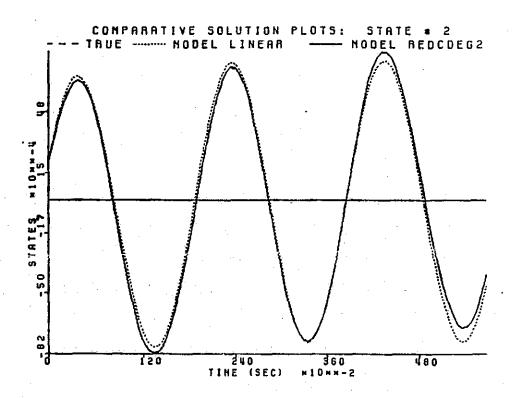


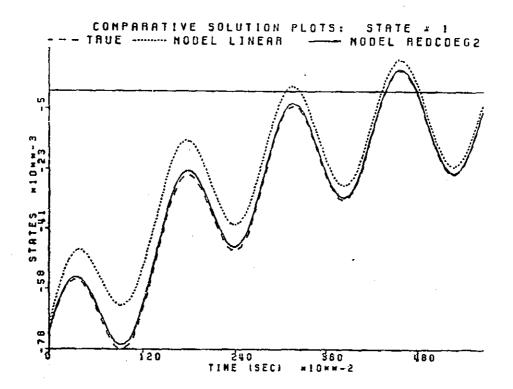
Figure 5.2.3 Simulation Number 6 of Table 5.2.1

The second secon

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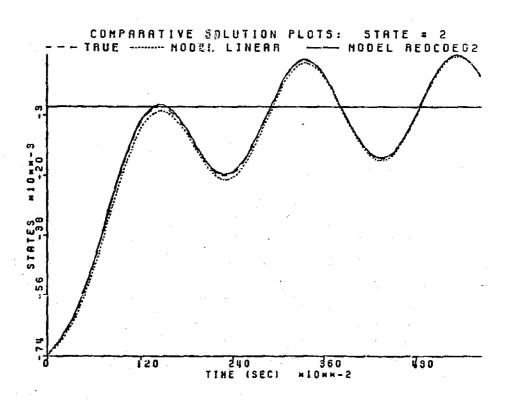


Figure 5.2.4 Simulation Number 15 of Table 5.2.2

FRORLEM SUMMARY CONFIGURATION: TRUE-LIMEAR, MODEL: # OF SINTES: 3 # OF TERMS IN MODEL 1: 16 # DEGREE OF AFFOXIMATION: 2 # SIMULATION WITH COSINE

St	INITIAL CONGI	TIONS	AMPLITUD	ES	FREQUENCIES		ERROR RAZIOS	
-23545-67-89-01-235-45-5	0.5000E-01 0.5000E-0 0.5000E-01 0.5000E-0 0.5000E-01 0.5000E-0 0.5000E-01 0.5000E-0 0.5000E-01 0.5000E-0 0.5000E-01 0.5000E-0 0.5000E-01 0.5000E-0 5000E-015000E-0 5000E-015000E-0 5000E-015000E-0 5000E-015000E-0 5000E-015000E-0 5000E-015000E-0 5000E-015000E-0	1 0.500E-01 1 0.500E-01 1 0.500E-01 1 0.500E-01 1 0.500E-01 1 0.500E-01 1 0.500E-01 1 -500E-01 1 -500E-01 1 -500E-01 1 -500E-01 1 -500E-01 1 -500E-01 1 -500E-01		0.050 0.050 -0.050 -0.050 -0.050 0.050 0.050 0.050 -0.050 -0.050	1.00 2.00 1.00 2.00 2.00 2.00 0.50 1.00 1.50 1.50 1.00 0.50 1.00 2.00 1.00 2.00 2.00 2.00 0.50 1.00 1.50 1.00 2.00 1.00 1.00 2.00 2.00 0.50 1.00 0.50 1.50 1.00 0.50 1.50 2.00 1.00 2.00 2.00 1.00 0.50 1.00 1.50 1.50 0.50 1.00	0.7637E-03 0.9997E-04 0.3218E-03 0.1279E-02 0.2859E-02 0.2052E-02 0.1139E-02 0.1743E-01 0.3096E-01 0.7710E-60 0.1457E-01 0.8380E-02 0.63327E-02 0.1750E-01	0.1131E-01 0.2249E-02 0.1303E-01 0.1974E-01 0.2857E-02 0.2811E-02 0.4268E-02 0.3458E-01 0.474E-02 0.3445E-01 0.6994E-01 0.2242E-01 0.2242E-01 0.244E-01	0.6534E-02 0.5413E-02 0.4332E-02 0.17892-01 0.1011E-01 0.9392E-02 0.3846E-01 0.2673E-01 0.2031E-01 0.4192E-01 0.2163E-01 0.2163E-01 0.5762E-01

Table 5.2.3 Simulation Table for Linear Model versus Second Degree Reduced Model

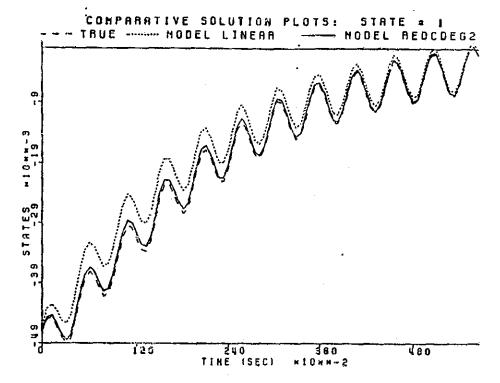
S t	INITIAL CONDIT	CONS	AHPLITU	DES	FREQUENCTES		ERROR RATIOS	
12345 67 B9 0112	0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01	0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01	0.050 0.05 0.050 0.05 0.050 0.05 0.050 0.05 0.050 0.05 0.050 0.05 0.050 0.07 0.050 0.07 0.050 0.07	0 0.050 0 0.050 0 0.050 0 0.050 0 0.050 0 0.025 0 0.025 0 0.025 0 0.025	0.00 0.25 0.50 9.50 0.00 0.25 0.25 0.50 0.00 1.00 0.55 0.25 0.50 0.25 1.00 0.00 0.25 0.50 0.50 0.25 0.50 0.50 0.25 0.50 0.50 0.00 0.00 0.25 1.00 0.00 0.25 1.00 0.55 0.50 0.25 1.00	0.4668E-03 0.2041E-01 0.7300E-02 0.4319E-03 0.1455E-02 0.4849E-03 0.1058E-02 0.8773E-01 0.2047E-02 0.9068E-03 0.2984E-02	0.2337E-01 0.3175E-01 0.4695E-01 0.1147E-01 0.1215E-01 0.2027E-01 0.7085E-01 0.1174 0.1754E-01 0.2663E-01 0.4477E-01	0.1146E-01 0.5587E-02 1.038 0.1388E-01 0.3617E-01 0.2464E-02 0.4997E-02 0.7417E-01 0.6591E-02 0.7651E-02 0.7181E-03

Table 5.2.4 Simulation Table for Linear Model versus Second Degree Reduced Model

Sŧ	INITIAL CONDITION	ONS	AMPLI: JUDES	FREQUENCIES	· .	ERROR RATIOS	
1234547890123454789012	0.5000E-01	0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.1000 -1000	0.075 -0.025 -0.050 0.075 -0.025 -0.050 0.075 -0.025 -0.050 0.075 -0.025 -0.050 -0.050 -0.025 -0.075 -0.050 -0.025 -0.075 0.050 -0.025 -0.075 0.050 -0.025 -0.075 0.050 -0.025 -0.050 0.025 0.075 0.050 0.025 0.075 0.050 0.025 0.075 0.050 0.075 -0.025 -0.050 0.075 -0.025 -0.055 0.050 -0.025 -0.075 0.050 -0.025 -0.075	0.00 0.00 0.00 0.02 0.02 0.02 0.05 9.05 0.10 0.10 0.10 0.10 0.02 0.02 0.02 0.05 0.05 0.05 0.10 0.10 0.10 0.00 0.66 0.00 0.02 0.02 0.02 0.05 0.05 0.05 0.10 0.10 0.10 0.00 0.00 0.00 0.02 0.02 0.02 0.05 0.05 0.05 0.10 0.10 0.10 0.02 0.02 0.02 0.05 0.05 0.05 0.10 0.10 0.10 0.02 0.02 0.02 0.05 0.05 0.05 0.10 0.10 0.10 0.02 0.02 0.02 0.05 0.05 0.05 0.05 0.05 0.05 0.05 0.05 0.05 0.05 0.05 0.05	0.1508E-03 0.1760E-03 0.1763E-03 0.1053E-02 0.2579 0.5782E-01 0.8089E-02 0.4316E-01 0.2966E-01 0.1200E-02 0.3750E-02 0.3760E-02 0.3721E-02 0.3721E-02 0.4025E-01 0.4025E-01 0.4025E-01 0.4025E-01 0.4025E-01	0.1754E-01 0.8649E-02 0.123E-01 0.5245 0.5597E-01 0.6644E-02 0.3525E-01 0.3025E-01 0.3724E-01 0.3724E-01 0.3549E-01 0.320E-01 3.549 0.9060 0.1830 0.1830 0.1830 0.1840E-01 0.1465E-01 0.9864E-02	0.3399 0.1457 0.4224 0.6259 0.7814 0.7137 1.418 0.5143E-01 0.3360E-01 0.3347E-01 0.4469 0.4789 0.6239 0.7801 0.8639 0.8639 0.8639 0.8639 0.9858 0.3277E-01 0.1175E-01 0.1934E-01 0.5811E-01

Table 5.2.5 Low Frequency Table for Linear Model versus Second Degree Reduced Model

ORIGINAL PAGE IS OF POOR QUALITY



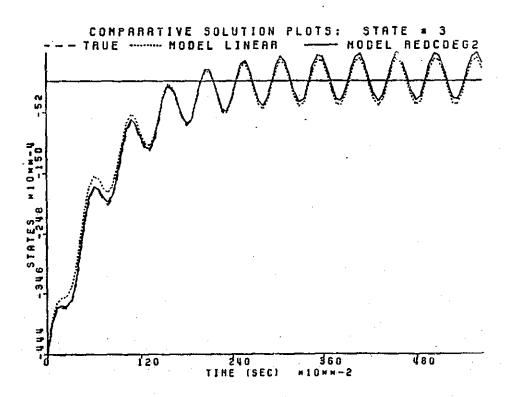
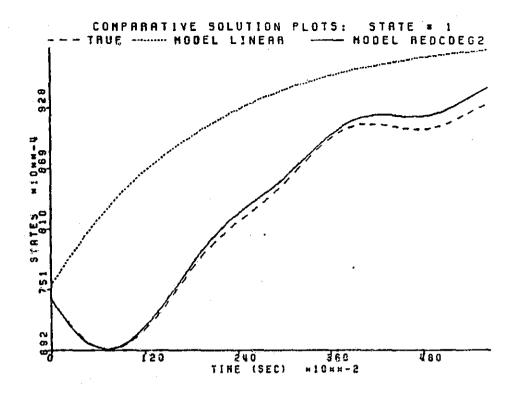


Figure 5.2.5 Simulation Number 10 of Table 5.2.3



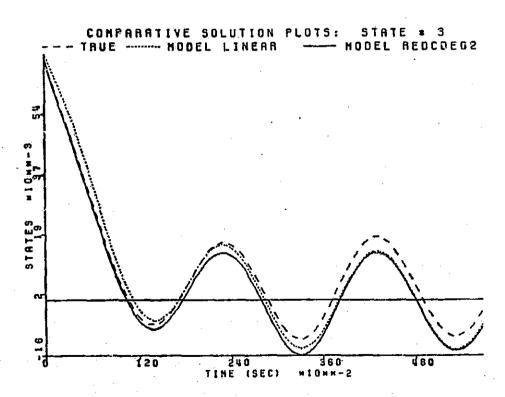
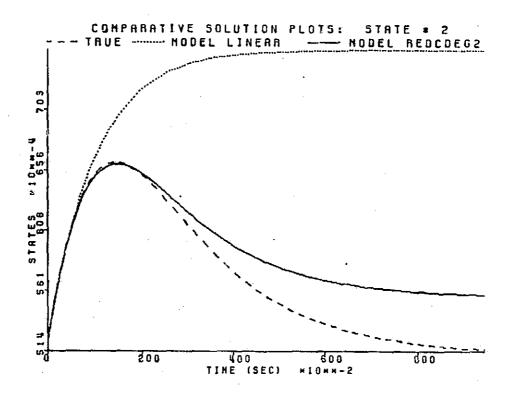
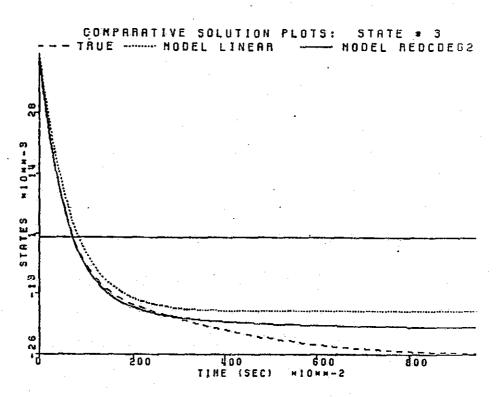


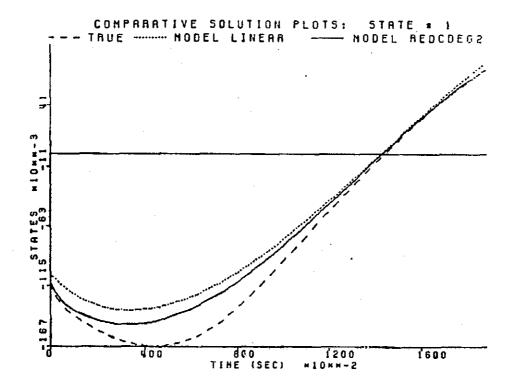
Figure 5.2.6 Simulation Number 3 of Table 5.2.4





STATE OF THE STATE

Figure 5.2.7 Simulation Number 1 of Table 5.2.5



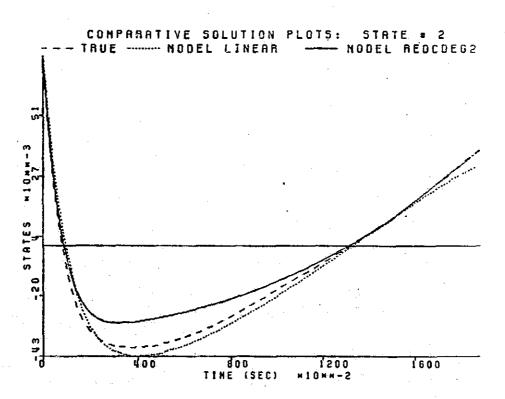


Figure 5.2.8 Simulation Number 16 of Table 5.2.5

PROBLEM SUMMARY
COMFIGURATION: TRUE MODEL1, MODEL2

OF STATES: 3

OF TERMS IN MODEL 1: 27
DEGREE OF APPROXIMATION: 2

OF TERMS IN MODEL 2: 16
DEGREE OF APPROXIMATION: 2

STMULATION WITH COSINE

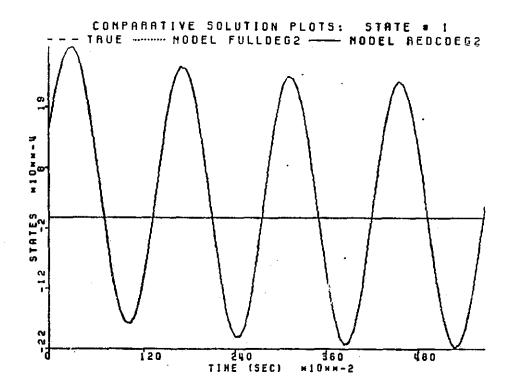
ORIGINAL PAGE IS OF POOR QUALITY

Sŧ	INITIAL CONDITI	ONS	ANGLITUDES	FREQUENCIES		ERROR RATIO	S
12345678901123456789012234	0.1000E-02 0.5000E-02	0.1000E-02 0.1000E-02 0.1000E-02 0.1000E-02 0.1000E-02 0.1000E-02 0.1000E-02 1000E-02 1000E-02 1000E-02 1000E-02 1000E-02 0.5000E-02 0.5000E-02 0.5000E-02 0.5000E-02	0.010 -0.010 -0.010 -0.010 -0.010 0.010 0.050 0.050 -0.050 -0.050 0.050 -0.050 -0.075 -0.075 -0.075 0.025 0.025 0.075 0.015 0.050 -0.025 0.010 -0.010 -0.010 -0.010 -0.010 -0.010 -0.050 0.050 -0.050 -0.050 0.050 0.050 -0.050 0.055 0.055 -0.055 0.055 0.055 -0.055 0.055 0.055 -0.055 0.055 0.055 -0.055 0.055 0.055	0.50 0.60 0.70 0.60 0.70 0.60 0	1.006 1.039 1.335 0.7310 1.105 0.9179 0.4727 0.5525 1.339 0.9802 1.339 0.9181 0.9181 1.128 0.9181 1.421 1.252 1.253 0.7664 1.054 0.9054 0.9054 0.9054 0.9054	0.5139 2.517 0.2341 0.6318 0.56318 0.56318 0.5705 0.6354 2.150 0.6385 0.6385 0.5625 0.2252 0.5741 1.603 1.155 1.2799 0.6887 0.5909 0.5909	0.9057 1.742 0.4745 0.3957 0.5476 0.1459 1.445 1.848 0.9311 1.787 0.4400 0.5489 0.1456 1.865 5.172 7.107 0.3715 0.2552 0.1587 0.1587

Table 5.2.6 Simulation Table for Second Degree Full Model versus Second Degree Reduced Model

15000E-025000E-02 0.5000E-02 0.010 -0.010 -0.010 0.50 0.60 0.70	1.185 0.4755 0.3688 0.9711 0.9145 0.6242 1.298 0.1848 0.3612
2 \$-5000E-02 -5000E-02 0.5000E-02 -0.010 -0.010 0.010 0.50 0.60 0.70 3 =-5000E-02 -5000E-02 0.5000E-02 -0.050 0.050 0.050 0.50 0.60 0.70 4 =-5000E-02 -5000E-02 0.5000E-02 -0.050 0.050 0.050 0.50 0.60 0.70 5 =-5000E-02 -5000E-02 0.5000E-02 -0.075 -0.075 -0.075 0.50 0.60 0.70 6 =-5000E-02 -5000E-02 0.5000E-02 -0.075 -0.075 0.050 0.50 0.60 0.70 7 =-5000E-02 -5000E-02 0.5000E-02 -0.075 -0.075 0.050 0.50 0.60 0.70 8 =-5000E-02 -5000E-02 0.5000E-02 0.075 0.050 0.050 0.50 0.60 0.70 9 =-7500E-01 -7500E-01 -7500E-01 -0.010 -0.010 0.010 0.50 0.60 0.70 10 =-7500E-01 -7500E-01 -7500E-01 -0.010 -0.010 0.010 0.50 0.60 0.70 11 =-7500E-01 -7500E-01 -7500E-01 -0.050 0.050 0.050 0.50 0.60 0.70 12 =-7500E-01 -7500E-01 -7500E-01 -0.050 0.050 0.050 0.50 0.60 0.70 13 =-7500E-01 -7500E-01 -7500E-01 -0.050 0.050 0.050 0.50 0.60 0.70 14 =-7500E-01 -7500E-01 -7500E-01 -0.050 0.050 0.050 0.50 0.60 0.70 15 =-7500E-01 -7500E-01 -7500E-01 -0.050 0.050 0.050 0.050 0.60 0.70 16 =-7500E-01 -7500E-01 -7500E-01 -0.050 0.050 0.050 0.050 0.60 0.70 17 0.7500E-01 -7500E-01 -7500E-01 0.050 0.050 0.050 0.050 0.050 0.60 0.70 18 0.7500E-01 -7500E-01 -7500E-01 0.050 0.050 0.050 0.050 0.60 0.70 19 0.7500E-01 -7500E-01 -7500E-01 0.050 0.050 0.050 0.050 0.60 0.70 19 0.7500E-01 -7500E-01 -7500E-01 0.050 0.050 0.050 0.050 0.60 0.70 19 0.7500E-01 -7500E-01 -7500E-01 0.050 0.050 0.050 0.050 0.60 0.70 19 0.7500E-01 -7500E-01 -7500E-01 0.050 0.050 0.050 0.050 0.60 0.70 19 0.7500E-01 -7500E-01 -7500E-01 0.050 0.050 0.050 0.50 0.60 0.70 19 0.7500E-01 -7500E-01 -7500E-01 0.050 0.050 0.050 0.50 0.60 0.70 20 0.7500E-01 -7500E-01 -7500E-01 0.050 0.050 0.050 0.50 0.60 0.70 21 0.7500E-01 -7500E-01 -7500E-01 0.050 0.050 0.050 0.50 0.60 0.70 22 0.7500E-01 -7500E-01 -7500E-01 0.050 0.050 0.050 0.50 0.60 0.70 23 0.7500E-01 0.7500E-01 -7500E-01 0.050 0.050 0.050 0.50 0.60 0.70 24 0.7500E-01 0.7500E-01 -7500E-01 0.050 0.050 0.050 0.50 0.60 0.70 25 -7500E-01 0.7500E-01 -7500E-01 0.050 0.050 0.050 0.50 0.60 0.70 26 -7500E-01 0.7500E-01 0.7500E-01 0.050 0.050 0.050 0.50 0.60 0.70	0.7292 0.5173 0.2126 1.066 0.4213 0.4112 0.9098 0.3041 - 0.1510 0.6557 0.5255 0.6750 0.5734 1.314 0.6442 1.065 0.5965 0.6772 1.065 0.5965 0.6772 1.074 0.6169 0.6201 0.9956 0.6737 0.6169 1.087 0.6393 0.6971 0.9984 0.6693 0.5352 1.168 0.6626 0.5683 0.9729 0.6243 0.6442 2.570 33.18 10.60 2.593 33.67 9.474 3.901 6.3855 5.753 2.203 6.294 7.673 1.641 2.863 6.225 2.113 2.930 12.22 2.1950 19.35 5.564 3.383 10.48 5.724 0.6615 0.7081 0.86722-02 0.6680 0.7347 0.9461E-02 0.7528 0.6559 0.823E-01 0.6860 0.7347 0.9461E-02 0.7528 0.6559 0.8231E-01 0.8610 0.4370 0.2356 0.8610 0.4370 0.2356

Table 5.2.7 Simulation Table for Second Degree Full Model versus Second Degree Reduced Model



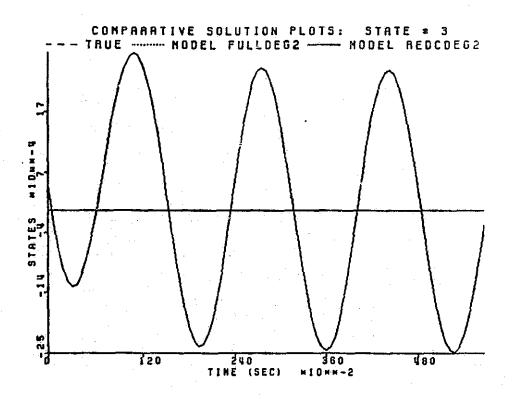
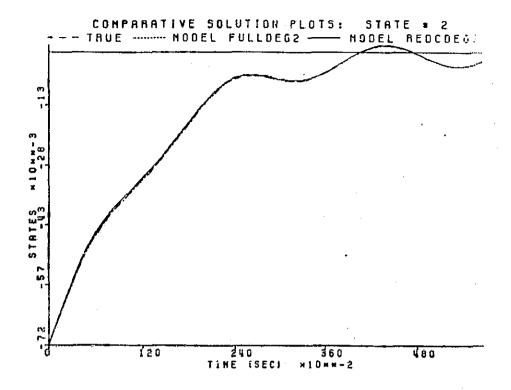


Figure 5.2.9 Simulation Number 2 of Table 5.2.6

1...

No. of Contract of

П



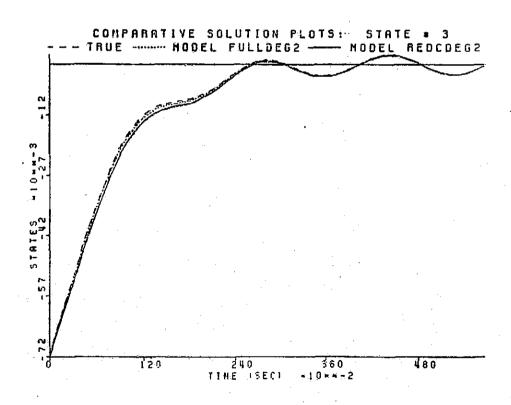


Figure 5.2.10 Simulation Number 17 of Table 5.2.7

Sŧ	IN	ITIAL CONDITI	OHS	AH	PLITUGES	FREQUENCIES		ERROR RATI	05
12345678901123456	0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 5000E-01 5000E-01 5000E-01 5000E-01 5000E-01	0.5000E-01 0.5000E-02 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 5000E-01 5000E-01 5000E-01 5000E-01 5000E-01 5000E-01	0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 5000E-01 5000E-01 5000E-01 5000E-01 5000E-01 5000E-01	0.050 0.050 0.050 0.055 -0.075 -0.075 -0.375 0.050 0.050 0.050 -0.075 -0.075	0.050 0.050 0.050 0.050 0.050 0.050 0.050 0.050 0.025 -0.050 0.025 -0.050 0.025 -0.050 0.025 0.050 0.050 0.050 0.050 0.050 0.050 0.050 0.050 0.050 0.025 -0.050 0.025 -0.050	1.00 2.00 1.00 2.00 2.00 2.00 0.50 1.00 1.50 1.50 1.00 0.50 1.00 2.00 1.00 2.00 2.00 2.00 1.50 1.00 0.50 1.00 2.00 1.50 1.00 2.00 1.00 0.50 1.00 1.50 1.50 1.00 0.50 1.00 2.00 1.50 1.00 2.00 1.50	1,934 0,2240 1,296 1,727 1,193 1,134 1,215 1,162 1,162 1,162 1,162 1,162 1,163	1.705 1.844 1.616 1.433 2.044 1.702 1.702 0.4919 0.4525 0.5461 0.5495B 0.5194 0.5222	1.375 1.174 1.460 1.579 1.579 2.093 0.4672 0.5042 0.5184 0.5184 0.4765

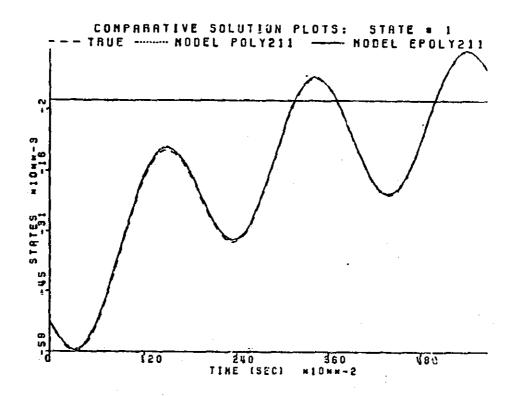
Table 5.2.8 Simulation Table for Second Degree Full Model versus Second Degree Reduced Model

2 0,7500E-01 0,7500E-01 0,7500E-01 0,050 0,050 0,050 0,050 0,050 0,00 0,23 1,043 1,476 0,9911 3 0,7500E-01 0,7500E-01 0,7500E-01 0,050 0,050 0,050 0,050 0,050 0,00 1,143 0,8409 2,006 4 0,7500E-01 0,7500E-01 0,7500E-01 0,050 0,050 0,050 0,050 0,23 1,00 0,50 1,229 1,374 1,437 5 0,7500E-01 0,7500E-01 0,7500E-01 0,050 0,050 0,050 1,00 0,50 0,25 1,167 1,168 1,689 6 0,7500E-01 0,7500E-01 0,7500E-01 0,050 0,050 0,050 0,050 0,050 1,169 1,408 1,111 7 0,7500E-01 0,7500E-01 0,7500E-01 0,050 0,075 0,025 0,00 0,25 0,50 0,00 0,050 1,080 1,368 2,778 8 0,7500E-01 0,7500E-01 0,7500E-01 0,050 0,075 0,025 0,00 0,025 0,50 0,008 1,368 2,778 9 0,7500E-01 0,7500E-01 0,7500E-01 0,050 0,075 0,025 0,00 0,02 1,080 1,935 1,140 9 0,7500E-01 0,7500E-01 0,7500E-01 0,050 0,075 0,025 0,50 0,00 0,02 1,080 1,935 1,140 9 0,7500E-01 0,7500E-01 0,7500E-01 0,050 0,075 0,025 0,50 0,00 0,02 1,080 1,935 1,140 9 0,7500E-01 0,7500E-01 0,7500E-01 0,050 0,075 0,025 0,25 0,50 0,00 0,9867 1,278 2,955 10 0,7500E-01 0,7500E-01 0,7500E-01 0,050 0,075 0,025 0,25 0,50 0,50 0,9861 1,376 1,404	Sŧ	INITIAL CONDITIONS		AMPLITUDES		FREQUENCIES		ERROR RATIOS			
12 0.7500E-01 0.7500E-01 0.7500E-01 0.050 0.075 0.025 0.50 0.25 1.00 0.9951 1.414 0.4005	1234567B91011	0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01	0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01	0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01 0.7500E-01	0.050 0.050 0.050 0.050 0.050 0.050 0.050 0.050	0.050 0.050 0.050 0.050 0.050 0.075 0.075 0.075	0.050 0.050 0.050 0.050 0.055 0.025 0.025 0.025	0.50 0.00 0.25 0.25 0.50 0.00 0.25 1.00 0.55 1.00 0.50 0.23 0.50 0.25 1.00 0.00 0.25 0.00 0.50 0.00 0.02 0.25 0.50 0.00 0.25 1.00 0.50	1.043 1.143 1.229 1.167 1.169 0.6098 1.080 0.9867 0.9601 1.074	1.476 0.8409 1.374 1.188 1.409 1.368 1.935 1.278 1.376	2.006 1.437 1.689 1.111 2.778 1.140 2.955

Table 5.2.9 Simulation Table for Second Degree Full Model versus Second Degree Reduced Model

59	INITIAL CONDITI	ONS	AMPLITUDES	FREQUENCIES	. 8	RROR RATIOS	
1234567B91011234567B91201123456786786786786786786786786786786786786786	0.5006E-01 0.5000E-01	0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.1000 -1000 -1000 -1000 -1000 -1000 -1000 -1000 -1000 -1000 -1000 -1000 -1000	0.075 -0.025 -0.050 0.075 -0.025 -0.050 0.075 -0.025 -0.050 0.075 -0.025 -0.050 0.075 -0.025 -0.075 -0.050 -0.025 -0.075 -0.050 -0.025 -0.075 0.050 0.025 0.075 0.050 0.025 0.075 0.050 0.025 0.075 0.050 0.025 0.075 0.050 0.075 -0.025 -0.050 0.075 -0.025 -0.050 0.075 -0.025 -0.050 0.075 -0.025 -0.050 0.075 -0.025 -0.050 0.075 -0.025 -0.050 0.075 -0.025 -0.050 0.075 -0.025 -0.050 0.075 0.025 -0.075 0.025 0.075 0.050 0.025 0.075 0.050 0.025 0.075 0.050 0.025 0.075 0.050	0.00 0.00 0.00 0.02 0.02 0.02 0.05 0.05 0.05 0.10 0.10 0.10 0.02 0.02 0.02 0.05 0.05 0.05 0.10 0.10 0.10 0.00 0.00 0.00 0.02 0.02 0.05 0.10 0.10 0.10 0.00 0.00 0.00 0.02 0.02 0.02 0.05 0.05 0.05 0.10 0.10 0.10 0.02 0.02 0.02 0.05 0.05 0.05 0.10 0.10 0.10 0.02 0.02 0.02 0.05 0.05 0.05 0.10 0.10 0.10 0.00 0.00 0.00 0.02 0.02 0.02 0.05 0.05 0.05 0.10 0.10 0.10	0.2240E-01 0.1406 0.2674 0.8874 0.9950 1.997 1.328 1.351 1.175 0.5226E-01 0.8891E-01 0.8517 0.2607 0.8700 0.9243 0.7724 1.038 0.7724	0.9726 1.099 1.099 1.143 0.9328 0.9328 0.9789 1.440 1.155 0.7468 0.7468 0.9651 0.9814 0.9788 0.9653 0.9789 0.9789 0.9783 0.9783	0.3866 0.1816 1.402 3.751 0.6507 0.5907 1.337 1.241 0.9409 0.6745 0.3995 0.3994 0.3995 0.3994 0.67350 0.6784 0.678

Table 5.2.10 Low Frequency Table for Second Degree Full Model versus Second Degree Reduced Model



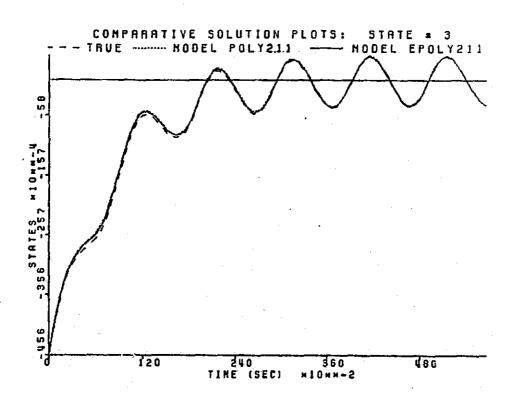
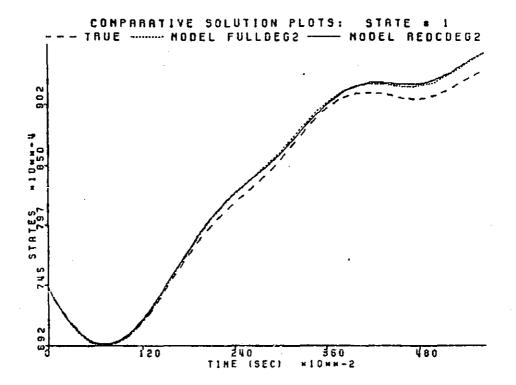


Figure 5.2.11 Simulation Number 16 of Table 5.2.8



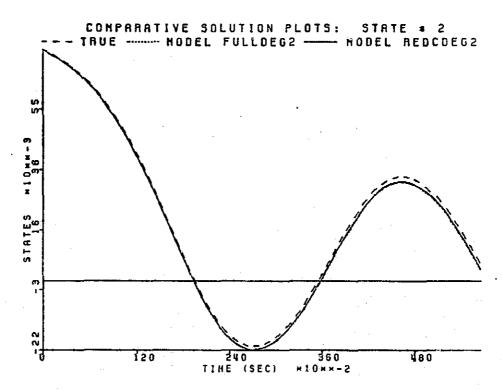
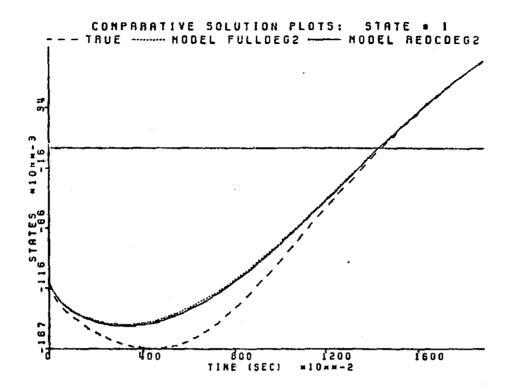


Figure 5.2.12 Simulation Number 9 of Table 5.2.9



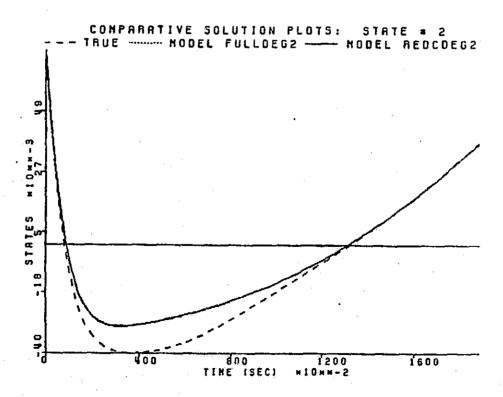
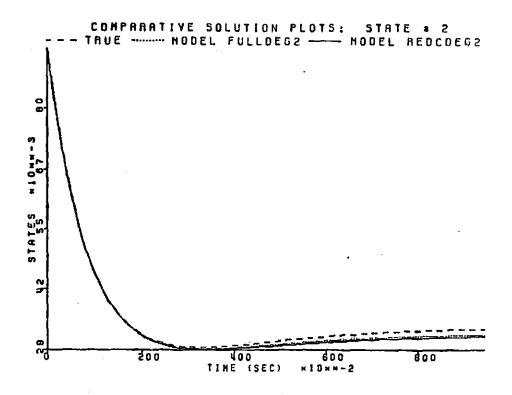


Figure 5.2.13 Simulation Number 16 of Table 5.2.10

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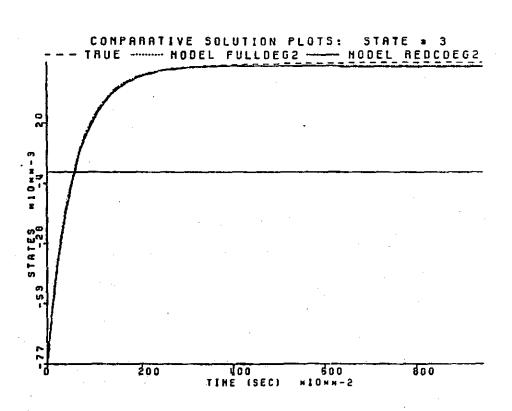


Figure 5.2.14 Simulation Number 19 of Table 5.2.10

5.3 THREE STATE, TWO INPUT SYSTEM

In this section the degree two approximation of a three state, two input system is studied. The system of three differential equations is again chosen to have the origin as a point of stable equilibrium from which it will be perturbed. The set of three equations is

$$\dot{x}1 = -x_1 + \cosh(x_1) + \sinh(x_2) - u_1e^{-u_2} + u_2$$

$$+ x_1x_2x_3 - e^{-x_3} \cosh(x_2) + 5x_1u_1u_2$$

$$- 2u_1^2 + x_2x_3,$$

$$\dot{x}_2 = -\frac{1}{2}x_2 + 6x_1^2 - 2e^{-x_3} - \frac{2}{x_2x_3} + x_1u_2e^{-x_3}$$

$$- \sinh(x_2) - \sinh(4x_3) + 2u_1 + 3u_1u_2 + 2,$$

$$\dot{x}_3 = -2x_3 + 5x_3u_2 - 2\cosh(3u_1) + \cosh(x_1)$$

$$+ \cosh(x_3) + x_1u_2 - \frac{1}{2}u_2 + \frac{2}{u_2}.$$

The second degree expansion contains 20 terms corresponding to the products

An acceptable model was identified at the initial conditions (0.01, 0.04, -0.02). And the input excitations are

$$u_1(t) = 0.025\sin(2\pi t \cdot \phi_1) + 0.025\sin(2\pi t \cdot 2\phi_1),$$

 $u_2(t) = -0.03\sin(2\pi t \cdot \phi_2) - 0.03\sin(2\pi t \cdot 2\phi_2).$

Where (ϕ_1,ϕ_2) was determined by local optimization to be (1.9941, 1.5518) from the starting frequencies (2.0, 1.6). This model and the corresponding reduction test appear in Figure 5.3.1.

ENTER INITIAL CONDITIONS FOR THE 3 STATES: 0.0100 0.0400 -0.0200 0.0100

ENTER SAMPLING PARAMETERS: APPROXIMATION DEGREE # SAMPLE POINTS INTERVAL BETWEEN SAMPLES INTEGRATION STEPSIZE

0.0400

100

SHOULD THE EXCITATION BE SINUSOIDAL OR COSINUSOIDAL?[5/C]:

0.0050

IDENTIFICATION WILL BE DONE WITH SINES.

ENTER THE NUMBER OF (CO)SINUSOIDS PER INPUT:

ENTER THE

ENTER THE 2 FREQUENCY WEIGHTS: 1.9941 1.5518

SHOULD THE IDENTIFICATION BE PERFORMED USING DATA THAT IS REDUCED?[Y/N]:

ENTER OPTIMIZATION OPTION:

1 - NO OPTIMIZATIONS
2 - AMPLITURES ONLY
3 - FREQUENCIES ONLY
4 - INITIAL CONDITIONS ONLY
5 - AMPLITURES AND FREQUENCIES
6 - AMPLITURES AND INITIAL CONDITIONS
7 - FREQUENCIES AND INITIAL CONDITIONS
8 - AMPLITURES, FREQUENCIES, AND INITIAL CONDITIONS

DO YOU WISH TO NORMALIZE THE DATA?[Y/N]:

THE MATRIX OF SAMPLED MONOMIAL TERMS HAS 20 ROWS AND 100 COLUMNS.

NUMBER OF TIMES COST FUNCTION WAS EVALUATED:

WOULD YOU LIKE TO SEE PLOTS OF THE INPUTS? IF SO, LINE UP THE CARRIAGE.

TSO SPEAKEASY III PI+ 8:14 PM FEBRUARY 22, 1984 :_SIZE=500;GET IDENT;IDENT;QUIT

EXECUTION STARTED

FARTITION NUMBER 1. PRTITION (A 3 BY 3 ARRAY)

Figure 5.3.1a Full Second Degree Model and Reduction Test (First Pass)

```
OF POOR QUALITY,
       -.99992 .99926 1.0047
-.0072833 -1.4999 -2.0127
-.0022769 .0019061 -1.9879
     ...WITH EIGENVALUES:
VALUES (A VECTOR WITH 3 COMPONENTS)
-1.9676 -1.5126 -1.0075
    PARTITION NUMBER 2.
FRITION (A 3 BY 2 ARRAY)
-.99886 .99926
1.9965 .0011641
4.1191E-5 -.49957
   PARTITION NUMBER 3.
FRIITION (A 3 BY 6 ARRAY)
.47926 .060162 -.28906
-.043156 -1.0172
-.44303
                                                                  --44635
--24907
                                                                                         1.2155
                                                                                                            -- 14853
                                                                                       -1.416
                                                                                                           -3.8262
                                            -,44303
                                                                  .0076769 .0016053 .83891
   PARTITION NUMBER 4.
PRITITION (A 3 BY 6 ARRAY)
.021255 -.046279 .025311
.060328 1.04 .04484
                                                                        -6.1793E-4 .069515
-.13033 .19197
-.012593 .05819
                                                                                                                      ~,022372
                                                                                                                      -.37207
        .0082058
                              1.009
                                                     .010457
                                                                                                  +05819
                                                                                                                       4.9249
  PARTITION NUMBER 5.
PRITITION (A 3 BY 3 ARRAY)
-1.9896 1.0341 .0023039
.0063837 -.0042395 -.0059681
-8.9742 3.753E-4 1.0006
 $ (A 20 COMPONENT ARRAY)

9.7773 6.9733 6.7771 4.317

1.5996 1.3777 1.023 .91586

.36988 .22983 .072894 .13248

THE MAXIMUM SINGULAR VALUE:

MAX = 9.7773

THE MINIMUM NONZERO SINGULAR VALUE:
                                                                            3,7318 3,5834
,81555 ,6472
                                                                                                  3.5834
                                                                                                                     3,0317
                                                                                                                                       1.8302
                                                                                                                     .50885
  MIN = .072894
  RTIO = 134.13
 WANT TO TRY REDUCTION TEST?CY/NJ: Y
 FULL SYSTEM ERROR:
FULL STATE ERRORS:
                                           2375,2090
229,7720 750,5278
                                                                                                1394.9093
 TERM: X1. COLUMN #: 1. CHANGE IN SYSTEM ERROR: -76.4391 CHANGE IN STATE ERRORS: -78.0076
                                                                                      1,9494
                                                                                                               -0.3809
                                      COLUMN #: 2,
ROR: 281,7762
RORS: 90,6889
TERM: X2. COLUM
CHANGE IN SYSTEM ERROR:
CHANGE IN STATE ERRORS:
                                                                                 191.0099
                                                                                                                 0.0774
TERM:
TERM: X3. COLUMN #: 3. CHANGE IN SYSTEM ERROR: -81.4371 CHANGE IN STATE ERRORS: -66.0698
                                                      -66.0678
                                                                               -166,4523
                                                                                                            151.0850
```

1

Figure 5.3.1b Full Second Degree Model and Reduction Test (First Pass)

			Ma	
TERM! CHANCE IN CHANGE IN	U1. COLUMN SYSTEM ERROR: STATE ERRORS:	-5,8238	4.3847	-0.0008
TERM: CHANGE IN CHANGE IN	U2. COLUMN SYSTEM ERROR: STATE ERRORS:	#; -20.3200 -3.1729	-0.0723	-22,1175
TERM: CHANGE IN CHANGE IN	X1.X1. COLUMN SYSTEM ERROR: STATE ERRORS:	*: -576.5861 37.3129	-679.8017	65,9026
TERM: CHANGE IN CHANGE IN	X1.X2. COLUMN SYSTEM ERROR: STATE ERRORS:	\$: 7. 3.2733 2.3766	3.7719	-2.8752
	X1.X3. COLUMN SYSTEM ERROR: STATE ERRORS:	8: 8. -52.8012 11.4518	-81.3662	17,1132
TERM: CHANGE IN CHANGE IN	X2.X2. COLUMN SYSTEM ERROR: STATE ERRORS:	\$: 9. -10.0474 -21.7582	11,6903	0.0185
TERM: CHANGE IN CHANGE IN	X2.X3. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 10. -90.7143 -49.0060	-41.7053	-0.0030
TERM: CHANGE IN CHANGE IN	X3.X3. COLUMN SYSTEN ERROR: STATE ERRORS:	#: 11. 207.7640 -6.7590	205,9084	8.6146
TERM: CHANGE IN CHANGE IN	X1.U1 COLUMN SYSTEM ERROR: STATE ERRORS:	#: 12. -1.5119 0.3419	- i.7045	-0.1493
TERM: CHANGE IN CHANGE IN	X1.U2. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 13. 19.4381 0.5890	-33,6165	43.4656
TERM: CHANGE IN CHANGE IN	X2.U1. COLUMN SYSTEM ERROR: STATE ERRORS:	*: 14. -0.2625 0.6154	-0.8328	-0.0450
TERM: CHANGE IN	X2.U2. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 15. 2.4908		-0.3592
TERM:	X3.U1. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 16.		
TERM!	X3.U2. COLUMN SYSTEM ERROR: STATE ERRORS:	± ± .17.		-45,3831
TERM: CHANGE IN	U1.U1. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 18. -1282.1752		-1127.7482
TEDM!	U1.U2. COLUMN SYSTEM ERROR: STATE ERRORS:	.a. 10		•
TERM:	U2.U2. COLUMN SYSTEM ERROR: STATE ERRORS:	#: 20.		104.8133

Figure 5.3.1c Full Second Degree Model and Reduction Test (First Pass)

DO YOU WANT TO DISCARD ANY TERMS AND RE-OPTIMIZETCY/N3: Y
HOW MANY TERMS WILL BE KEPT? 15

ENTER THE COLUMN NUMBERS OF COLUMNS WHICH ARE TO BE KEPT.
ENTRY OF COLUMN NUMBERS WILL TAKE PLACE IN BLOCKS OF 10.
WHEN ASKED, ENTER A AS AN ARRAY WITH A MAXIMUM OF TEN ELEMENTS.
A = ARRAY(10:1:2:3:4:5:6:8:9:10:11)

A = ARRAY(5:13,17,18,19,20)

MANUAL MODE SPACE USED 82 K NOW, 95 K PEAK, SIZE 500 K DO YOU WISH TO SAVE THIS MODEL?[Y/N]:N DO YOU WISH TO IDENTIFY ANOTHER MODEL?[Y/N]:Y

Figure 5.3.1d Full Second Degree Model and Reduction Test (First Pass)

From the reduction test, there are 5 columns of the model which appear to be relatively insignificant. They are

These terms were removed and another optimization over frequencies was performed. This time, given the starting frequency set (1.994, 1.5518), the routine converged to frequencies (2.005, 1.553). This model and the reduction test appear in Figure 5.3.2. The test indicates that there remain no terms whose effect is negligible. From this point, reduction can proceed with careful choice of any term to be removed. For the purpose of demonstration, though, we accept the 5 term reduction as the final one. This is a 25% reduction and again all linear terms were kept.

With the model identified, simulation remains. As in the previous section, for the purpose of presentation only two of the three states will be plotted for any given simulation. Those chosen are generally a worst case and a typical plot. The first five tables and their corresponding plots are simulations of the exact linear model versus the reduced second degree model. Again the ratio is the mean square error of the reduced model over the mean square error of the linear approximation. In this case, the exact linear model appears

$$\mathbf{L}_{10} = \begin{bmatrix} -1 & 1 & 1 \\ 0 & -1.5 & -2 \\ 0 & 0 & -2 \end{bmatrix}, \quad \mathbf{L}_{01} = \begin{bmatrix} -1 & 1 \\ 2 & 0 \\ 0 & -0.5 \end{bmatrix}.$$

Table 5.3.1 and Table 5.3.2 show the simulation data for moderately sized excitations. Table 5.3.3 stretches the amplitudes out further than the previous tables. In Table 5.3.4 random excitations were chosen. The initial conditions were chosen in the range (-0.075, 0.075). And the cosinusoidal input

```
ENTER INITIAL CONDITIONS FOR THE 3 STATES: 9.0100 0.0400 -0.0200
ENTER SAMPLING PARAMETERS:
APPROXIMATION DEGREE
                ¥ SAMPLE POINTS
INTERVAL BETWEEN SAMPLES
INTEGRATION STEPSIZE
        100
                     0.0400
                                     0.0050
SHOULD THE EXCITATION BE SINUSOIDAL OR COSINUSOIDAL?ES/CI:
INENTIFICATION WILL BE DONE WITH SINES.
ENTER THE NUMBER OF (CO)SINUSOIDS PER INPUT:
 NTER THE 4 INITIAL INPUT AMPLITUDES: 0.24999999999999996E-01
ENTER THE
ENTER THE 2 FREQUENCY WEIGHTS:
                           1,5518
SHOULD THE IDENTIFICATION BE PERFORMED USING DATA THAT IS REDUCED?[Y/N]:
ENTER OPTIMIZATION OPTION:

1 - NO OPTIMIZATIONS
2 - AMPLITUDES ONLY
3 - FREQUENCIES ONLY
4 - INITIAL CONDITIONS ONLY
              5 - AMPLITUDES AND FREQUENCIES
6 - AMPLITUDES AND INITIAL CONDITIONS
7 - FREQUENCIES AND INITIAL CONDITIONS
              8 - AMPLITUDES, FREQUENCIES, AND INITIAL CONDITIONS
ENTER THE TRACE OF THE COVARIANCE MATRIX: 0.99999999999999954E-06
ENTER THE OPTIMIZATION METHOD: 1,2,3 OR 4
ENTER # OF SIGNIFICANT DIGITS FOR CONVERGENCE, MAXIMUM # OF FUNCTION CALLS, AND THE ZXMIN OFTION # (0,1,2,3)
2 200 2
  10 CALLS OF FCN....
THE PARAMETERS:
0.20028408+01 0.15507558+01
THE CONDITION NUMBER: 0.7228+08
OBJECTIVE FUNCTION VALUE: 0.95568+03
20 CALLS OF FCN....
THE PARAMETERS:
_0.2004800D+01 0.155
0.20048000+01 0.1552782D+01
THE CONDITION NUMBER: 0.721D+08
QBJECTIVE FUNCTION VALUE: 0.9551D+03
30 CALLS OF FCN....
THE PARAMETERS:
0.2004831D+01 0.1552813D+01
THE CONDITION NUMBER: 0.720D+08
ORJECTIVE FUNCTION VALUE: 0.95510+03
     Figure 5.3.2a
                                 Reduced Second Degree Model and Reduction
```

Test (Second Pass)

```
40 CALLS OF FCM.....
0.2064810D+01 0.1552793D+01
THE CONDITION NUMBER: 0.721D+08
OBJECTIVE FUNCTION VALUE: 0.9551D+03
CONVERGENCE WAS ACHIEVED AND NO ERRORS OCCURRED.
DO YOU WISH TO NORMALIZE THE DATA?[Y/N]:
THE MATRIX OF SAMPLED MONOMIAL TERMS
      15 ROWS AND 100 COLUMNS.
NUMBER OF TIMES COST FUNCTION WAS EVALUATED:
                                                                                  47
THE OPTIMUM FREQUENCIES ARE: FREQ(1)= 2.005 FREQ(2)=
                                FREQ(2)= 1.553
                                                                FREQ(3) = 0.000
WOULD YOU LIKE TO SEE PLOTS OF THE INPUTS? IF SO, LINE UP THE CARRIAGE.
 MESSAGE SUMMARY: MESSAGE NUMBER - COUNT
TSO SPEAKEASY III PI+ 9:26 PM FEBRUARY 22, 1984 :_SIZE=500;GET IDENT;IDENT;QUIT
EXECUTION STARTED COLSKEPT (A 15 COMPONENT ARRAY) 1 2 3 4 5 6 8 9
                                                     10 11 13 17
                                                                               18
FARTITION NUMBER 1.
FRITTION (A 3 BY 3 ARRAY)
-.9995B .99946 1.0
-.008155B -1.5014 -2.0
                    9.6701E-4 -1.9909
  -,0025067
...WITH EIGENVALUES:
VALUES (A VECTOR WITH 3 COMPONENTS)
-1.9736 -1.5101 -1.0082
PARTITION NUMBER 2.
PRITITION (A 3 BY 2 ARRAY)
-.99869 .99925
1.9967 .0015374
 -7.3392E-6 -.49957
PARTITION NUMBER 3.
PRITITION (A 3 BY 6 ARRAY)
.4238 0 -.56573
6.0525 0 -1.1585
                              -.56573
-1.1585
                                                          1.2253
-.88992
                                            -.43827
```

FARTITION NUMBER 4.

.58504

Figure 5.3.2b Reduced Second Degree Model and Reduction Test (Second Pass)

.024684 .040922

.78135

-.10333

-.13797

```
PRTITION (A 3 BY 6 ARRAY)
0 -.040326 0 0 0 -.015627
0 1.0415 0 0 0 .061985
0 1.0091 0 0 0 4.9553
```

PARTITION NUMBER 5.
PRITITION (A 3 BY 3 ARRAY)
-1.9891 1.0348 .0031636
.0041551 -.0061485 -.0015756
-8.9747 -2.4192E-4 1.0005

S (A 15 COMPONENT ARRAY)
9.5482 6.7397 5.7757 4.2381 3.6701 3.5553 2.9761 1.2907 1.4226
.80701 .72912 .4787 .41589 .0802 .28158
THE MAXIMUM SINGULAR VALUE:
MAX = 9.5482
THE MINIMUM NONZERO SINGULAR VALUE:
MIN = .0802

RTIO = 119.06

WANT TO TRY REDUCTION TEST? LY/NJ: Y

FULL SYSTEM ERROR: 2360.2953 FULL STATE ERRORS: 227.9609	737.6836 1394.6509
TERM: X1. COLUMN #: 1. CHANGE IN SYSTEM ERROR: -75.24 CHANGE IN STATE ERRORS: -76.98	43
TERM: X2. COLUMN #: 2. CHANGE IN SYSTEM ERROR: 276.67 CHANGE IN STATE ERRORS: 89.90	12
TERM: X3. COLUMN #: 3. CHANGE IN SYSTEM ERROR: -73.48 CHANGE IN STATE ERRORS: -65.32	91
TERM: U1. COLUMN #: 4. CHANGE IN SYSTEM ERROR: -7.98 CHANGE IN STATE ERRORS: -10.62	37
TERM: U2. COLUMN #: 5. CHANGE IN SYSTEM ERROR: -26.47 CHANGE IN STATE ERRORS: -4.36	66
TERM: X1.X1. COLUMN #: 6. CHANGE IN SYSTEM ERROR: -566.23 CHANGE IN STATE ERRORS: 32.17	88
TERM: X1.X3. COLUMN #: 8. CHANGE IN SYSTEM ERROR: -62.60 CHANGE IN STATE ERRORS: 22.61	14
TERM: X2.X2. COLUMN #: 9. CHANGE IN SYSTEM ERROR: -16.51 CHANGE IN STATE ERRORS: -21.14	49
TERM: X2.X3. COLUMN #: 10. CHANGE IN SYSTEM ERROR: -76.01 CHANGE IN STATE ERRORS: -48.63	34
TERM: X3.X3. COLUMN #: 11.	

Figure 5.3.2c Reduced Second Degree Model and Reduction Test (Second Pass)

	/ SYSTEM ERROR: ! STATE ERRORS:	204.0818 -9.3087	204.6296	6.7610
TERM: CHANGE IN CHANGE IN	X1.U2. COLUMN SYSTEM ERROR: STATE ERRORS:	13. 14.3135 0.5307	-30.3188	44.1017
	X3.U2. COLUMN N SYSTEM ERROR: N STATE ERRORS:	1: 17. -43.2243 -0.3695	1.5486	-44.4034
TERM: CHANGE IN CHANGE IN		#: 18. 1290.6057 -152.7579	-0.5303	-1137.3176
	U1.U2. COLUMN N SYSTEM ERROR: N STATE ERRORS:	#: 19. -16.7355 -16.8697	0.1380	-0.0037
	U2.U2. COLUMN SYSTEM ERROR: STATE ERRORS:	\$: 20. 104.6273 0.2396	0.2167	104.1710

DO YOU WANT TO DISCARD ANY TERMS AND RE-OPTIMIZE?[Y/N]: N

MANUAL MODE SPACE USED 68 K NOW, 80 K PEAK, SIZE 500 K NO YOU WISH TO SAVE THIS MODELTLY/NJ:N
10 YOU WISH TO IDENTIFY ANOTHER MODELTLY/NJ:N

Figure 5.3.2d Reduced Second Degree Model and Reduction Test (Second Pass)

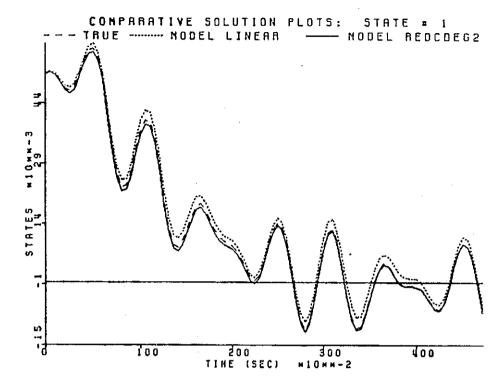
58	INITIAL CONDITIONS		AMPLITUDES	FREQUENCIES	ERROR RATIOS	
779 75 67 B 9-0 1019 45 67 87 0 1019 45 67 B 9-0 1019 45	0.5000E-01	-01 0.5008E-01 -01 0.5000E-01 -01 0.5000E-01 -01 0.5000E-01 -01 0.5000E-01 -01 0.5000E-01 -01 0.5000E-01 -01 -5000E-01 -01 0.5000E-01	0.0750 0.0750 0.0750 -0.0750 -0.0750 0.0750 -0.0750 0.0750 0.1500 0.1500 -0.1500 0.1500 -0.1500 -0.1500 -0.1500 0.1500 -0.1500 0.1500 -0.0750 0.0750 -0.0750 0.0750 -0.0750 0.1500 -0.1500 0.1500	1.50 1.50	0.9749E-01 0.3371E-02 0.7 0.1073 0.3553E-02 0.7 0.1116E-01 0.4384E-02 0.3 0.1117E-01 0.4484E-02 0.3 0.1117E-01 0.4282E-02 0.3 0.1115E-01 0.4282E-02 0.3 0.1759E-03 0.1386F-03 0.6 0.1759E-02 0.1289E-03 0.6 0.1759E-02 0.1517E-03 0.6 0.145E-02 0.1517E-03 0.6 0.145E-02 0.5129E-02 0.3 0.9048E-02 0.5048E-02 0.3 0.9048E-02 0.5048E-02 0.3 0.9049E-02 0.5048E-02 0.3 0.8951E-02 0.5038E-02 0.3 0.8951E-02 0.129E-02 0.6 0.6629E-02 0.129E-02 0.6 0.7677E-01 0.4528E-02 0.3 0.1594E-01 0.4728E-02 0.3 0.1594E-01 0.4728E-02 0.3 0.1594E-01 0.4728E-02 0.3 0.4325E-02 0.1718E-02 0.3 0.4325E-02 0.1718E-02 0.3 0.4325E-02 0.1718E-02 0.3 0.5555E-02 0.1718E-02 0.3 0.5555E-02 0.1718E-02 0.3 0.5555E-02 0.1914E-02 0.3 0.1799E-01 0.6099E-02 0.3 0.1799E-01 0.6099E-02 0.3 0.1797E-01 0.6099E-02 0.3 0.1677E-01 0.6178E-02 0.3 0.1677E-01 0.6178E-02 0.3	924E-04 737E-04 737E-04 663E-03 713E-04 663E-03 713E-03 714E-03 744E-03 747E-03 777E-03 777E-03 777E-04 777E-03 777E-04 777E-03 777E-04 777E-03 777E-04 777E-03 777E-04 777E-03 777E-04 777E-03

Table 5.3.1 Simulation Table for Linear Model versus Second Degree Reduced Model

	<i>f</i>		,		
Sŧ	INITIAL CONDITIO	DNS	AMPLITUDES	FREQUENCIES	EFROR RATIOS
123456789012114516789012134547890112		0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-01 0.5000E-015000E-015000E-015000E-015000E-015000E-01 0.5000E-01	0.0750	2.00 1.50 2.00 1.50	0.2436E-02 0.9744E-03 0.1277E-03 0.2378E-02 0.9892E-03 0.1805E-03 0.1817E-02 0.9766E-03 0.1805E-03 0.1817E-02 0.9766E-03 0.155E-03 0.1817E-03 0.1819E-01 0.4379E-02 0.3803E-03 0.1404E-01 0.4379E-02 0.3803E-03 0.1404E-01 0.4379E-02 0.3813E-03 0.1343E-01 0.4379E-02 0.3913E-03 0.1343E-01 0.4379E-02 0.3973E-04 0.1346E-01 0.1876E-02 0.4673E-04 0.1576E-04 0.1576E-04 0.1576E-04 0.1576E-04 0.1576E-04 0.1576E-01 0.1576E-02 0.4577E-03 0.1748E-01 0.5719E-02 0.3977E-03 0.1748E-01 0.5719E-02 0.3977E-03 0.1748E-01 0.5719E-02 0.3977E-03 0.1786E-01 0.1876E-02 0.3977E-03 0.1786E-01 0.1876E-02 0.3977E-03 0.1786E-01 0.1876E-02 0.3977E-03 0.1786E-01 0.1857E-03 0.2513E-03 0.1026E-01 0.1857E-03 0.2513E-03 0.1206E-01 0.1857E-03 0.2513E-03 0.2513E

Table 5.3.2 Simulation Table for Linear Model versus Second Degree Reduced Model

ORESEVAL PAREL TO OF POOR QUALITY



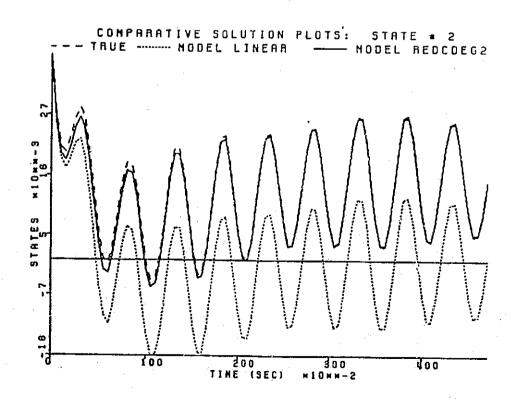
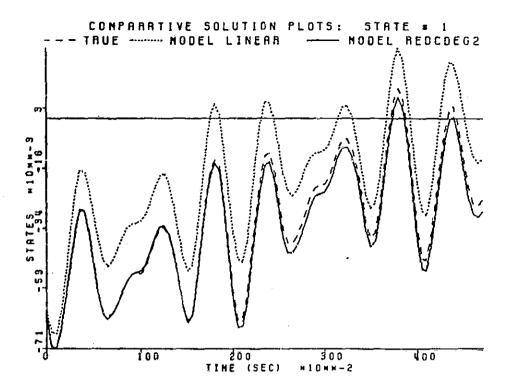


Figure 5.3.3 Simulation Number 4 of Table 5.3.1



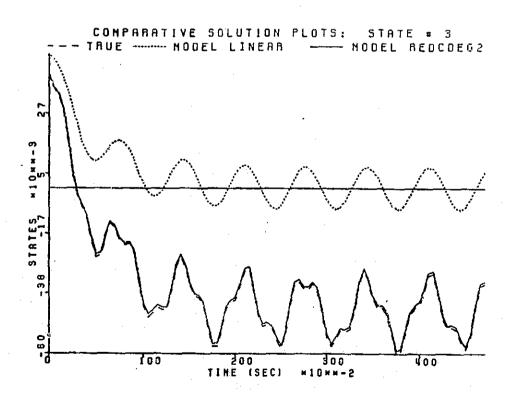


Figure 5.3.4 Simulation Number 22 of Table 5.3.2

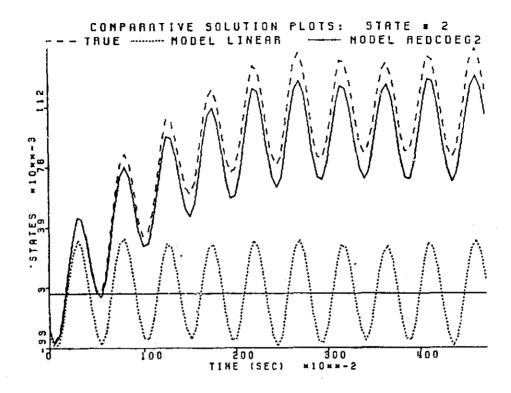
FROREM SUMMARY CONFIGURATION: TRUE-LINEAR-MODEL1 OF STATES: 3 OF INFUTS: 2 FOR FERNS IN MODEL 1: 15 DEGREE OF AFFACKIANTION: 2 STAULATION WITH COSINE

SI	INITIAL CONDITIONS		AMPLITUDES	FREQUENCIES	ERROR RATIOS	
27345-67890112H1453452853H2J345557859H	0.1000E-01 0.1000E-01	0.1000E-01	0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 1000E-01 1000E-01 1000E-01 1000E-01 1000E-01 1000E-01 1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 1000E-01 1000E-01 1000E-01 1000E-01 1000E-01 1000E-01 1000E-01 1000E-01 1000E-01 1000E-01 1000E-01 1000E-01 1000E-01	0.2009 0.2000 0.2000 -0.2009 -0.2000 -0.2009 -0.2000 -0.2000 0.3000 -0.3000 -0.3000 -0.3000 -0.3000 -0.3000 -0.3000 -0.2000 0.2000 -0.2000 -0.2000 -0.2000 -0.2000 -0.2000 -0.2000 -0.3000 -0.3000 0.3000 -0.3000 0.3000 -0.3000 0.3000 -0.3000 0.3000 -0.2000 -0.2000 0.2000 -0.2000 0.2000 -0.2000 0.2000 -0.2000 0.2000 -0.2000 0.2000 -0.2000 0.2000 -0.2000 0.2000 -0.2000 0.2000 -0.2000 0.2000 -0.2000 0.3000 0.3000 0.3000 0.3000 0.3000 0.3000 0.3000 0.3000 0.3000 0.3000 0.3000 0.3000 0.2000 -0.2000 0.2000 -0.3000 0.3000	2.10 1.60 2.10 1.60	0.3361E-01 0.1687E-01 0.111EE-02 0.350E-01 0.1739E-01 0.1117E-02 0.3516E-01 0.1697E-01 0.1117E-02 0.3565E-01 0.1739E-01 0.1117E-02 0.3565E-01 0.792F-01 0.566EE-02 0.792E-01 0.9442E-01 0.5616E-02 0.792E-01 0.9395E-01 0.5648E-02 0.792E-01 0.792E-01 0.5579E-02 0.7365E-01 0.737E-01 0.138E-02 0.3652E-01 0.792E-01 0.113E-02 0.3652E-01 0.792E-01 0.113E-02 0.3762E-01 0.174E-01 0.112E-02 0.3667E-01 0.1745E-01 0.112E-02 0.3667E-01 0.792E-01 0.564E-02 0.812E-01 0.9365E-01 0.564E-02 0.814E-01 0.9365E-01 0.564E-02 0.814E-01 0.936E-01 0.566E-02 0.814E-01 0.936E-01 0.568E-02 0.398E-01 0.157E-01 0.113E-02 0.398E-01 0.177E-01 0.113E-02 0.398F-01 0.177E-01 0.113E-02 0.398F-01 0.177E-01 0.559E-02 0.8367E-01 0.925E-01 0.568E-02 0.837E-01 0.925E-01 0.569E-02 0.337E-01 0.975E-01 0.113E-02 0.397E-01 0.975E-01 0.569E-02 0.838E-01 0.975E-01 0.569E-02 0.838E-01 0.975E-01 0.569E-02 0.838E-01 0.975E-01 0.569E-02 0.838E-01 0.975E-01 0.569E-02

Table 5.3.3 Simulation Table for Linear Model versus Second Degree Reduced Model

91	INITIAL COMPITIONS		AMPLITUDES	FREQUENCIES	ERFOR RATIOS
1234547896112345478990123352233535	3562E-016873E-016859E-016326E-01 0.7058E-012750E-013505E-01 0.1764E-013442E-01 0.2906E-01 0.3455E-016117E-01 0.3455E-016117E-02 0.3747E-01 0.7765E-01 0.6743E-01 0.2906E-01 0.6743E-01 0.2906E-01 0.6743E-01 0.3707E-024555E-013124E-01 0.6743E-01 0.3707E-024555E-013124E-01 0.507334E-01 0.3707E-024555E-013124E-01 0.5610E-01 0.3104E-015110E-02 0.7268E-015110E-02 0.7268E-01 0.5610E-01 0.3104E-01 0.7414E-01 0.6748E-01 0.7546E-01 0.3104E-01 0.7546E-01 0.3104E-01 0.7546E-01 0.3530E-01 0.5610E-01 0.3104E-01 0.7714E-012595E-01 0.3679E-01 0.3261E-01 0.3679E-01 0.3500E-01 0.3679E-01 0.3500E-01 0.3679E-01 0.3500E-01 0.3679E-01 0.3703E-01 0.3679E-01 0.3703E-01 0.3679E-01 0.3703E-01 0.4039E-01 0.4743E-01 0.4039E-01 0.4743E-01 0.4039E-01 0.4743E-01 0.4039E-01 0.4736E-01	0.2658E-022565E-01 0.6701E-013541E-013348E-013741E-01 0.3014E-01 0.3014E-01 0.3014E-01478E-01 0.3014E-01 0.3025E-01 0.3025E-01 0.3025E-01 0.3079E-01	0.1068 -0.2188 -0.0475 0.1321 0.0368 -0.1764 -0.2272 -0.0738 0.1093 0.0399 0.2190 -0.2008 0.0304 0.0827 -0.1400 0.1952 0.1417 -0.2474 -0.2217 -0.0448 0.2317 -0.0948 -0.0555 -0.0328 -0.1525 -0.1689 -0.0513 -0.1143 -0.2166 0.0507 -0.0745 -0.0818 -0.2284 -0.2463 0.2219 0.1643 0.0973 0.1462 -0.0492 -0.0861 0.0973 0.1462 -0.0492 -0.0861 0.0973 0.1462 -0.0374 0.1687 0.1942 -0.2352 0.1164 -0.2291 -0.2467 -0.1628 0.0490 0.2103 -0.1598 0.2165 0.1999 -0.2383 -0.1598 0.2165	2.72 3.43 2.72 1.08 1.72 1.09 1.76 1.76 1.76 1.76 1.76 1.77 1.78 1.78 1.79 1.76 1.77 1.78 1.79 1.70	0.3651E-02

Table 5.3.4 Simulation Table for Linear Model versus Second Degree Reduced Model



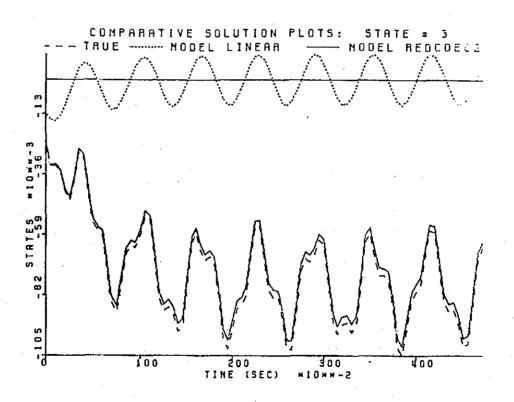
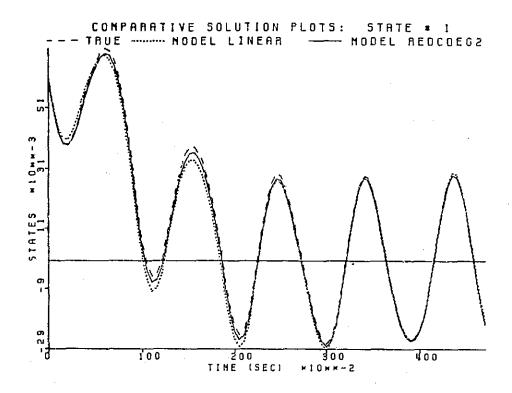


Figure 5.3.5 Simulation Number 11 of Table 5.3.3



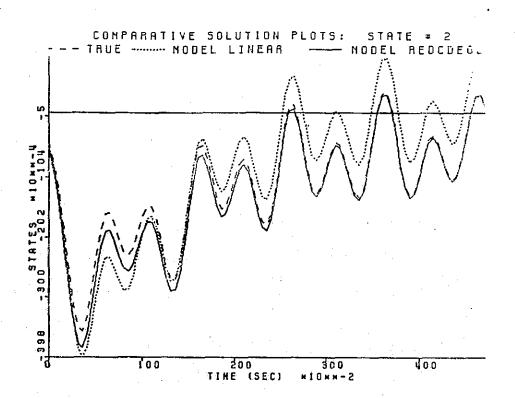
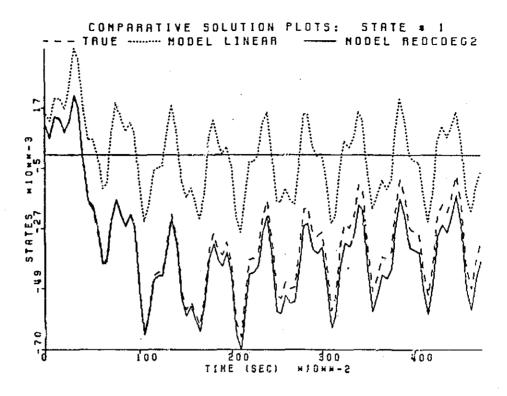


Figure 5.3.6 Simulation Number 3 of Table 5.3.4



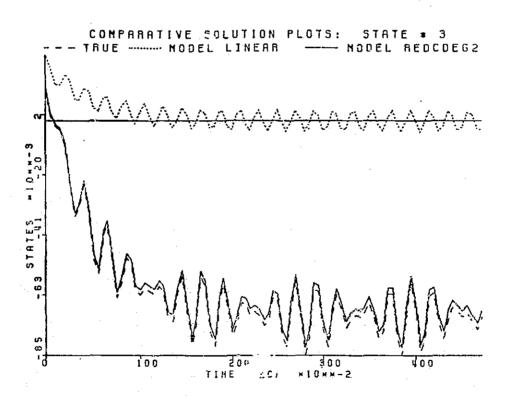


Figure 5.3.7 Simulation Number 27 of Table 5.3.4

amplitudes and frequencies were chosen in the ranges (-0.25, 0.25) and (1,5) respectively. Finally, Table 5.3.5 tests the d.c. and low frequency behavior of the models. In all cases, the reduced model outperforms the linear model.

As the final test of model degradation, the same simulations are performed for the full second degree model versus the reduced second degree model. Again in each of the five tables, the error ratio is the mean square error of the reduced model over the mean square error of the full model. In brief, Table 5.3.6 and Table 5.3.7 contain excitations of moderate size, while Table 5.3.8 moves the amplitudes out farther away from the origin. Table 5.3.9 is a table of randomly chosen excitations, and Table 5.3.10 is a low frequency test. Note in all cases, with the exception of the Table 5.3.10, the error ratios are near unity. This implies that almost no information was lost with the removal of the five terms. In Table 5.3.10, it appears that the third state has been hurt the most by the reduction, but Figure 5.3.14 shows that even a worst case example is not bad. All things considered, the reduction, although only five terms, has tested out well.

5.4 REMARKS

In this chapter three examples are studied to demonstrate the reduction method. For each example system, a reduction of the second degree model is carried out, and for the first example, the degree three approximation is also reduced. In each exercise the reduction method yields a smaller model while retaining the nonlinearities associated with the additional degree of approximation.

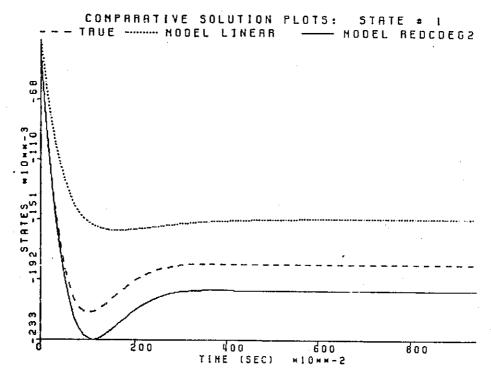
As a note on the computation involved in the reduction test, it is worth mentioning that the current version of the reduction test software is written

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12	INITIAL CO	CHOITIGHS	AMPLITUDES	FREQUENCIES	ERROR RATIOS	
	0.5000E-02	0E-02	9,0190 0,	0.0130010300103001030010300103001030010	0.5441E-02	0.90361-03 0.15595-03 0.75595-03 0.75595-03 0.35145-03 0.35145-03 0.35145-03 0.437345-03 0.137345-03 0.137345-03 0.137345-03 0.137345-03 0.137345-03 0.137345-03 0.137345-03 0.137345-03 0.137345-03 0.137345-03 0.137757-03

Table 5.3.5 Low Frequency Table for Linear Model versus Second Degree Reduced Model

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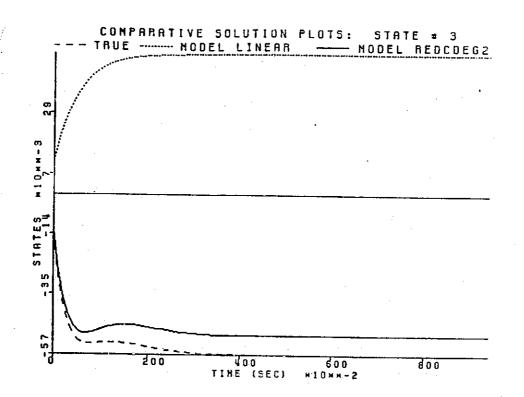
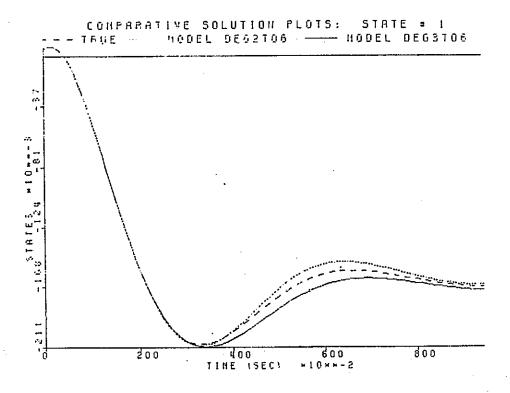


Figure 5.3.8 Simulation Number 10 of Table 5.3.5



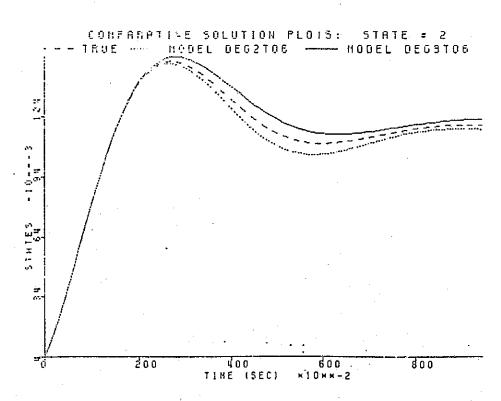
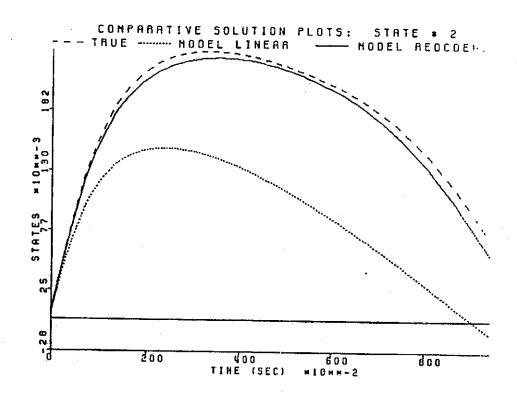


Figure A.33 S#5, Table A.18: Compare to Figure A.30



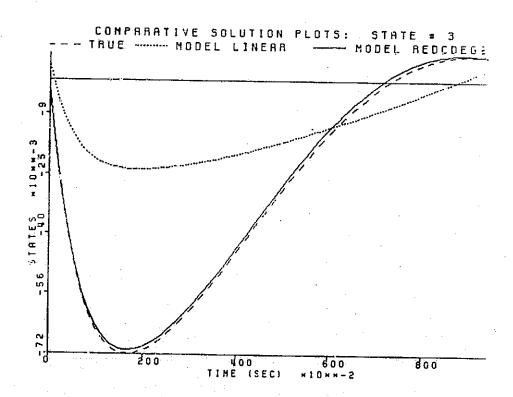


Figure 5.3.9 Simulation Number 33 of Table 5.3.5

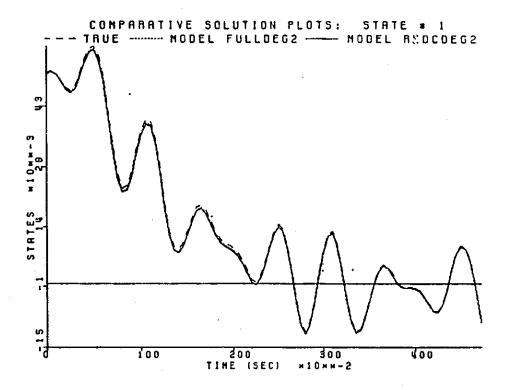
FAGELEM SURMARY CONFIGURATION: TRUE-MODELY-MODEL2 OF STATES: 3 OF TERMS IN MODEL 1: 20 DEGREE OF APPROXIMATION: 2: 15 DEGREE OF APPROXIMATION: 2 SIMULATION WITH COSINE

51	INITIAL CONDITIONS		DHS	AMPLITUDES	FREQUENCIES	÷	ERROR RATIOS	i
127454789011213451178781287875178788	0.5000E-01	0.500E-01	0.5900E-01 0.500E-01 0.500E-01 0.500E-01 0.500E-01 0.500E-01 0.500E-01500E-01500E-01500E-01500E-01500E-01 0.500E-01	0.0750 0.0750 0.0750 -0.0750 -0.0750 0.0750 -0.0750 0.0750 0.1550 0.15500 0.1550 -0.1590 -0.1500 -0.1590 -0.1500 -0.1590 -0.1500 -0.1590 0.0750 -0.0750 -0.0750 -0.0750 -0.1500 -0.1500 0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.0750 -0.0750 -0.0750 -0.0750 -0.0750 -0.0750 -0.0750 -0.0750 -0.0750 -0.0750 -0.0750 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.1500 -0.0750	2.00 1.50 2.00 1.50	1.434 1.439 1.452 1.530 1.452 1.527 1.6236 0.6764 0.8137 1.910 1.910 1.741 1.774 1.819 0.8137 1.774 1.774 0.8137 1.774 1.774 1.774 1.774	1.132 1.198 1.219 1.333 1.333 1.333 1.333 1.344 1.344 1.349 1.334 1.463 1.463 1.463 1.463 1.579 1.395 1.344 1.347 1.347 1.347 1.347 1.347 1.347 1.347 1.342 1.344 1.342 1.342 1.342 1.344 1.342 1.342 1.342 1.342 1.342 1.344 1.342 1.342 1.342 1.342 1.342 1.342 1.342 1.344 1.342 1.342 1.342 1.344 1.342 1.342 1.342 1.342 1.344 1.344 1.342 1.342 1.342 1.344 1.344 1.342 1.342 1.344 1.344 1.344 1.344 1.342 1.342 1.344	2.821 2.902 2.902 2.902 1.002 1.002 1.002 1.002 0.5545 0.5545 0.5545 0.5903 1.453 1.453 1.453 1.453 0.7024 0.56810 0.68810 0.68810 0.68810 0.68810 1.055

Table 5.3.6 Simulation Table for Second Degree Full Model versus Second Degree Reduced Model

58	INITIAL CONDITIONS		INITIAL CONDITIONS		AMPLITUDES	FREQUENCIES		ERROR RATIOS	•
123+567890112345478901237456789013	-500E-01 0.500E-01 -500E-01 -500E-01 0.500E-01 0.500E-01 -500E-01 0.500E-01 -500E-01 0.500E-01 0	0.5000E-01	0.0750	2.00 1.50 2.00 1.50	0.937377 0.937475 1.0174 1.017	0.6711 0.7129 0.7225 0.7225 1.357 1.353 1.353 1.8817 1.997 1.861 1.421 1.423 1.423 1.423 1.426 1.146 1.240 1.399 1.399 1.399 1.399 1.399 1.399 1.385 1.385 1.385 1.385 1.385 1.385	0.6049 0.6173 0.6173 0.5870 1.087 1.087 1.087 1.085 1.411 1.548 1.104 1.105 1.105 0.9980 0.9449 1.2243 1.2243 1.2243 1.237 1.337 1.338		

Table 5.3.7 Simulation Table for Second Degree Full Model versus Second Degree Reduced Model



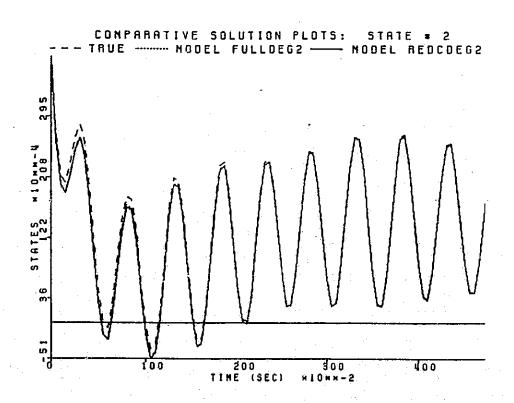
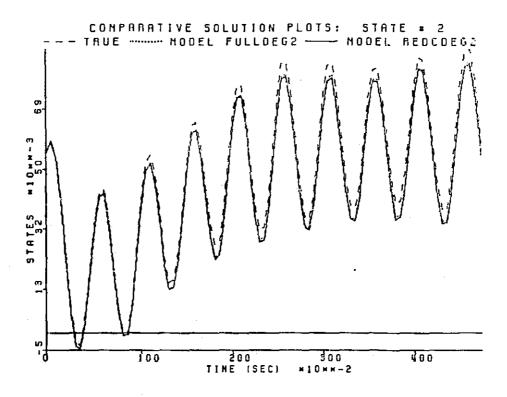


Figure 5.3.10 Simulation Number 4 of Table 5.3.6



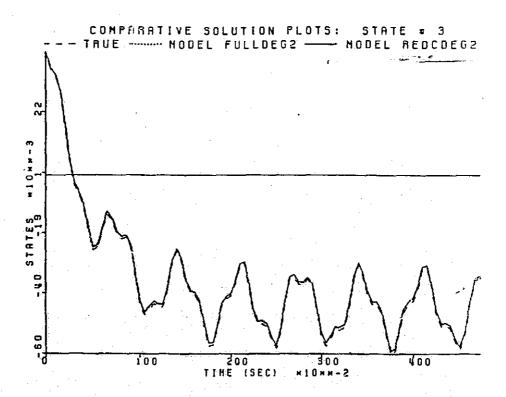


Figure 5.3.11 Simulation Number 5 of Table 5.3.7

ORIGINAL PAGE IN OF POOR QUALITY PROPLEM SUMMARY
CONFIGURATION: TRUE, MODEL1, MODEL2

& GF STATES: 3

& OF INPUTS: 2

& OF TERMS IN MODEL 1; 20

DEGREE OF APPROXIMATION: 2

PEGREE OF APPROXIMATION: 2

PEGREE OF APPROXIMATION: 2

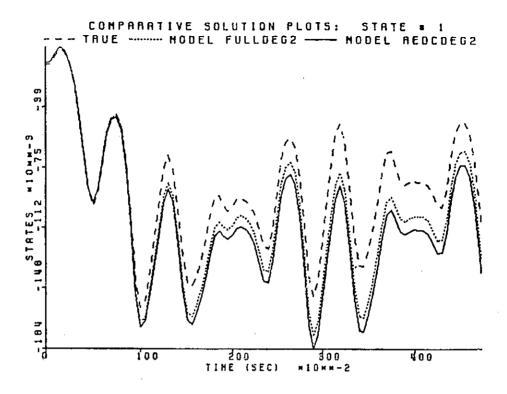
SIMULATION WITH COSINE

S#	THI	IAL CONDITI	DNS	AMPLITUDES	FREQUENCIES		ERROR RATIOS	
123456789011211415147892222222222222222	0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 -1000E-01	0.1000E-01 0.1000E-01	0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 -1000E-01 -1000E-01 -1000E-01 -1000E-01 -1000E-01 -1000E-01 -1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 0.1000E-01 -1000E-01 -1000E-01 -1000E-01 -1000E-01 -1000E-01 -1000E-01 -1000E-01 -1000E-01	0.2000 0.2000 0.2000 -0.2000 -0.2000 -0.2000 0.3000 -0.3000 0.3000 -0.3000 0.3000 -0.3000 0.2000 0.2000 0.2000 0.2000 0.2000 0.2000 0.2000 0.3000 -0.3000 -0.3000 0.3000 -0.3000 0.3000 -0.3000 0.3000 0.2000 0.2000 0.2000 0.2000 0.2000 0.3000	2.10 1.60 2.10 1.60	1.820 1.780 1.780 1.829 1.761 1.760 1.768 1.805 1.777 1.779 1.810 1.759 1.718 1.759 1.759 1.759 1.759 1.759 1.759 1.759 1.759 1.759 1.759 1.759 1.759 1.759 1.759 1.759	1.299 987 1.2999 1.3056 1.11554 1.11558 1.11558 1.11558 1.11558 1.11558 1.11558 1.11559 1.11559 1.11554 1.11559 1.11554 1.11559	1.1189972233832444111855744511993332441199971181111955714557111111111111111111111111111

Table 5.3.8 Simulation Table for Second Degree Full Model versus Second Degree Reduced Model

							•
5#	INITIAL CONDITI	CONS	amplitudes	FREQUENCIES		ERROR RATIOS	i
+23745-6789-01123345-5789-0123345-6789-03335-3345	3542E-016893E-016659E-016320E-01 0.7058E-012950E-024633E-021100E-013905E-01 0.7266E-01 0.4164E-027165E-01 0.3455E-01 0.3874E-01 0.3475E-01 0.2396E-025552E-01 0.3246E-01 0.7219E-01 0.2200E-016679E-01 0.2396E-024655E-014825E-01 0.5646E-01 0.3728E-024555E-016979E-017052E-013128E-013728E-026579E-017052E-01 0.3128E-017052E-01 0.3272E-01 0.3631E-01 0.3726E-017456E-01 0.3726E-017456E-01 0.3104E-017456E-01 0.3104E-01 0.7456E-01 0.3104E-01 0.7456E-01 0.3104E-01 0.7456E-01 0.3536E-01 0.3636E-01 0.3536E-01 0.3636E-01 0.3636E-01 0.7445E-01 0.3648E-01 0.7449E-01 0.3636E-01 0.36376E-013938E-01 0.3679E-013938E-01 0.3679E-013938E-01 0.3639E-013328E-01	0.2058E-02 2565E-01 0.6901E-01 3541E-01 43412E-01 4312E-01 0.3014E-01 0.2795E-01 0.3014E-01 0.4525E-01 4525E-01 0.3025E-01 0.3025E-01 0.3025E-01 0.3025E-01 0.3025E-01 0.4377E-01 0.4377E-01 0.4377E-01 0.4377E-01 0.4378E-01 0.4378E-01 0.4378E-01 0.4378E-01 0.4378E-01 0.4304E-01 0.4378E-01 0.4378E-01 0.4304E-01 0.4378E-01 0.4378E-01 0.4304E-01 0.4378E-01	0.1068 -0.2188 -0.0475 0.1321 0.0396 -0.1964 -0.2272 -0.0738 0.1093 0.0339 0.1193 0.0339 0.0304 0.0827 -0.1400 0.1952 0.1417 -0.2474 -0.2217 -0.0048 -0.0455 -0.0328 -0.0455 -0.0328 -0.0513 -0.0513 -0.2166 0.0120 -0.0513 0.0557 -0.0745 -0.0618 -0.2284 0.0120 -0.0513 0.0557 -0.0745 -0.0618 -0.2284 0.2463 0.0751 0.0494 0.1115 0.1214 -0.1877 -0.1140 0.0452 -0.1038 -0.0452 -0.1038 -0.0452 -0.1038 -0.0452 -0.1038 -0.0452 -0.1038 -0.0492 -0.0851 -0.0452 -0.1038	3,43 2,72 4,666 1,07 3,29 4,81 3,29 1,76 3,42 3,90 4,82 3,90 4,78 1,75 1,75 1,75 1,75 1,76 1,77 4,78 1,79 4,78 1,79 4,78 1,79 4,78 1,79 4,78 1,79 4,78 1,88 1,79 4,77 4,88 1,44 3,63 3,63 3,63 4,98 4,97 4,87 4,87 4,87 4,87 4,87 4,87 4,87 4,87	1.202 1.766 2.116 1.779 1.824 0.574 1.774 1.785 1.774 1.885 1.671 0.8501 1.8507	1.184 0.77579 1.783 1.259 1.574 0.5023 1.302 1.302 1.302 1.302 1.326 1.326 1.326 1.326 1.326 1.326 1.326 1.326 1.326 1.326 1.327 1.326 1.326 1.326 1.327 1.326 1.327 1.328 1.3	1.518 2.755 2.755 2.755 2.755 1.2172 1.192 0.35127 1.1267 1.1267 1.1263 1.1264 1.1264 1.1264 1.1264 1.1264 1.1264 1.1264 1.125 0.5957 2.382 1.168 1.16
31 32	4102E-015523E-01	1621E-02 0.3802E-01 0.6202E-01	-0.1399 0.2236 -0.1598 0.2165	1.25 3.95 4.97 3.22	1.539 1.877	1.168 1.345	1.271
33	7302F 01 0.6745E-01	9008E-02	0.1904 -0.1361	1.74 3.32	1.765	1,291	1,152

Table 5.3.9 Simulation Table for Second Degree Full Model versus Second Degree Reduced Model



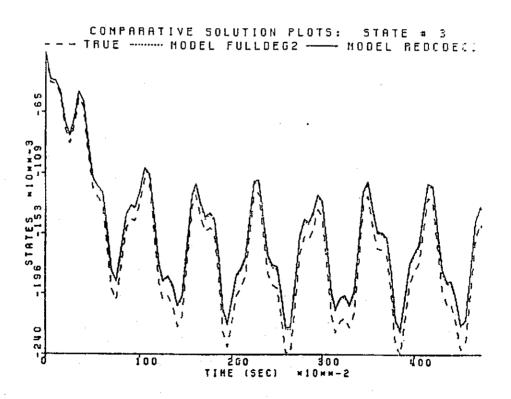
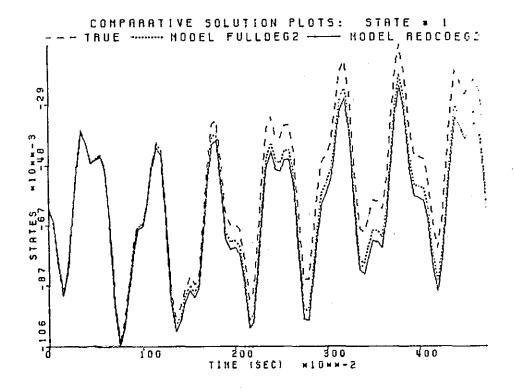


Figure 5.3.12 Simulation Number 21 of Table 5.3.8



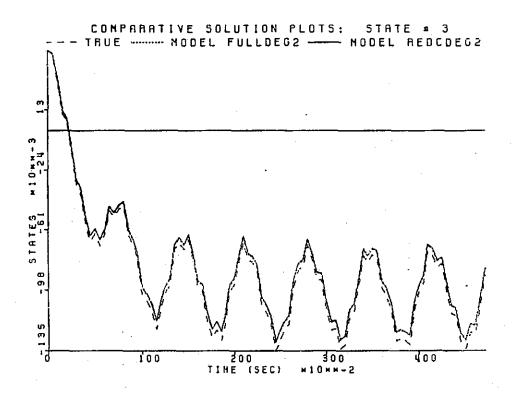
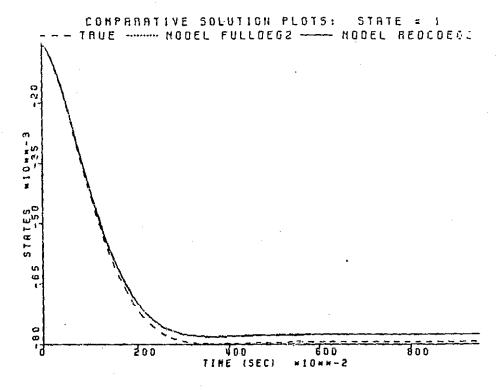


Figure 5.3.13 Simulation Number 18 of Table 5.3.9

FROSLEM SUMMARY CONFIGURATION: TRUE, KODEL1, MODEL2 OF STATES: 3 OF STATES: 2 OF SEEMS IN MODEL 1: 20 DEGREE OF AFFROXIMATION: 2 OF SEEMS IN MODEL 2: 15 DEGREE OF AFFROXIMATION: 2 SIMULATION WITH COSINE

1 0.500E-02 0.500E-02 0.500E-02 0.500E-02 0.0100 0.0100 0.010 0.01 1.221 0.5027 0.2013 2 0.500E-02 0.500E-02 0.500E-02 0.0300E-02 0.0100 0.0100 0.010 0.03 1.221 0.5027 0.7014 2 0.500E-02 0.500E-02 0.500E-02 0.500E-02 0.0100 0.0100 0.03 0.03 1.212 0.5022 0.7014 2 0.500E-02 0.500E-02 0.500E-02 0.500E-02 0.0500E-02 0.000E-02 0.500E-02 0.	50	INITIAL CONDITI	0HS	AHFLITUDES	FREQUENCIES		ERROR RATIOS	
47 -,100E-01 -,100E-01 -,100E-01 0,2000 -0,2000 0.01 0.01 0,E50 1,362 0,7270 48 -,100E-01 -,100E-01 0,2000 -0,2000 0.03 0.03 0.8577 1,345 0,7237	123456789011234567899123456789901233456789901233456789901233456	0.5006-02 0.5006-03 0.5006-02 0.5006-04 0.5006-04 0.5006-01 0.5006-01 0.1006-01	9.59906E-022 9.	6.0100 0.0100 0.0100 0.0100 0.0100 0.0100 0.0100 0.0100 0.0100 0.0100 0.0759 -0.0750 0.1000 0.1000 0.1600 0.1000 0.2000 -0.2000 0.2000 -0.2000 0.2000 -0.2000 0.2000 -0.0750 0.01000 0.0100 0.0100 0.0100 0.0100 0.0100 0.0100 0.0100 0.0100 0.01000 0.0100 0.0100 0.0100 0.0100 0.0100 0.0100 0.0100 0.0100 0.01000 0.0100	0.601 0.601	0.1.550044979448100880974944359595733449508809795495575575575579557381495088097955753444500880979549755755755755755755755755755755755755755	0.001.3344.5344.5345.5414.400.000.3344.5345.5414.400.000.000.000.000.000.000.000.000.	0.2835 0.4012 0.7934 0.805 0.1037 0.1037 0.1037 0.7287 0.7287 0.7287 0.7287 0.7287 0.7287 0.7287 0.7287 0.7287 0.7287 0.7287 0.7287 0.7287 0.7287 0.7339 0.7

Table 5.3.10 Low Frequency Table for Second Degree Full Model versus Second Degree Reduced Model



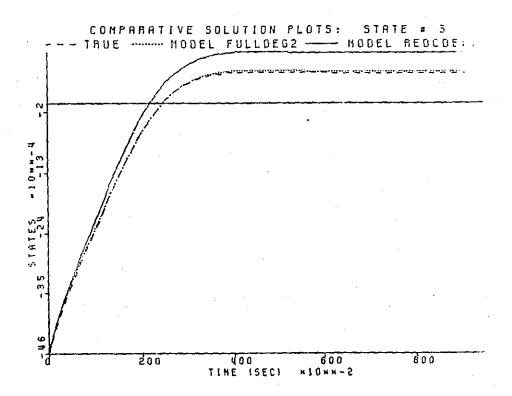
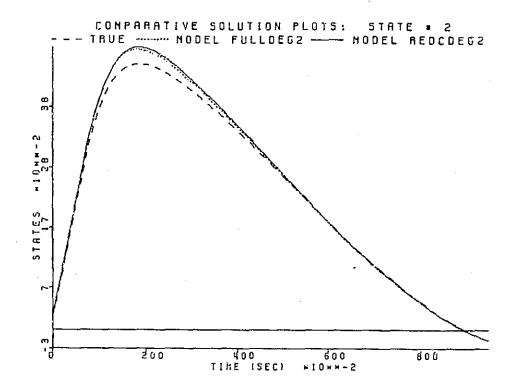


Figure 5.3.14 Simulation Number 16 of Table 5.3.10



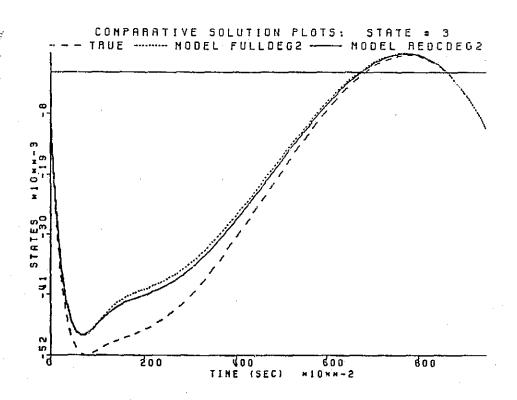


Figure 5.3.15 Simulation Number 36 of Table 5.3.10

in the FORTRAN based language SPEAKEASY. While SPEAKEASY makes the iterative elimination of rows and columns a trivial task, it is (being a very high level computer language) inherently slow. A reduction test about the size of those presented here, would take, on the average, about 5 minutes of cpu time. This time of execution might be shortened singificantly if future versions of reduction software are written directly in FORTRAN. This may even become necessary as model size increases, but for the initial efforts contained herein, the slow execution is accepted.

CHAPTER VI

CONCLUSIONS

The primary goal of this thesis was to initiate studies of and develop a methodology for the size reduction of the nonlinear models generated in the tensor framework. This goal was accomplished using an idea based on the error inherent in the model due to the least squares fit of the data. The iterative reduction scheme was developed primarily by logical argument and intuition.

Based on the reasoning, a loose step by step reduction methodology was formulated. Using the reduction technique, system identification and reduction were carried out for several examples. In all cases the results were good and demonstrated that a reduced model can be identified which retains influential higher order nonlinearities while discarding the insignificant ones in an effort to decrease the size of the model.

Based on the encouraging nature of the results documented here, further research into the following ideas is warranted. First, with respect to this particular reduction method, Desrochers has suggested that cost functions can be implemented to select the appropriate terms in the appropriate combinations so as to aid in the omitting of terms. Perhaps his optimization ideas may be modified to fit this particular problem formulation as was his idea for model reduction.

A second suggestion has to do with extending the model reduction technique to the layered model identification approach. This identification scheme has been studied in [14] and concerns itself with modelling the error in some model approximation with the addition of another degree of non-linearity. The outlook for this identification technique is promising.

The current reduction method could be modified to enable a layer by layer reduction. Or if an entire layered medel is to be reduced, provisions would have to be made to save the data from each previously modelled layer, as the current reduction method requires it for the error test. In either case, the application of the reduction test to the layered modelling approach is a problem that deserves study.

In sum, based on the findings of this thesis, further research is justified not only in the extension of the reduction method to other identification approaches, but also into the improvement and advancement of the current reduction scheme.

APPENDIX A [See Acknowledgments, page iii] NONLINEAR TENSOR MODELS: A CASE STUDY

A.1 INTRODUCTION

So often schemes for various procedures in nonlinear system theory such as parameter estimation, inverse systems, and so on, are discussed, whether it be in textbooks or technical papers, but are left void of any substantial convincing proof of validity or utility. Indeed, even some of the simplest of nonlinear systems present problems which require extensive illustration of a technique to produce an acceptable level of reliability. It is this proof-in-the-pudding spirit to which we adhere in this report to study the tensor parameterization approach.

Let us begin by presenting the nonlinear system to be modeled. Two coupled nonlinear differential equations given by

$$x_{1} = u_{2}\cosh(x_{1}x_{2}) - 3\sinh(x_{2}) - e \sinh(2x_{1})$$

$$u_{1}u_{2} \qquad u_{1}$$

$$x_{2} = \sinh(x_{2}) + e \qquad \sinh(x_{1}) - e \quad u_{1}\cosh(x_{1}^{2})$$
(A.1.1)

serve as our example system, where as usual x_1 and x_2 represent the states, x_1 and x_2 their time derivatives, and u_1 and u_2 the system inputs. In order to utilize the tensor parameterization modeling approach, the allowable input set must be bounded in some reasonable sense, particularly for low frequency cases. In fact, the system (A.1.1) exhibits finite escape for a relatively small step input in the first input channel. The L_{10} matrix (that is, the matrix of parameters associated with linear terms in x alone) of the linear approximation for (A.1.1),

$$\dot{x}_1 = u_2 - (2x_1 + 3x_2)
\dot{x}_2 = x_1 + x_2 - u_1,$$
(A.1.2)

has characteristic values

$$\lambda_{1} = -\frac{1}{2} \pm j \sqrt{3/2}, i = 1,2.$$
 (A.1.3)

Transient responses to step inputs are of five to six seconds in duration. Note further that the origin, (x,u) = 0, is a stable equilibrium point.

In the section to follow specifics relative to the actual identification scheme are discussed, and we comment on the extensive software capabilities which have been developed to facilitate these modeling studies. Following this are two sections illustrating a second degree approximation model and a third degree approximation model. Verification studies in the form of simulation tables and representative plots make up the bulk of these sections. In the final section we discuss some further considerations of these particular models, and dwell on possibilities for improved performance.

A.2 ALGORITHM AND SOFTWARE CONSIDERATIONS

The overall modeling procedure can be separated into three basic stages: problem formulation, model identification, and model verification. In the first stage, the system is formulated in the differential equation description; in the general case we assume this form of system description, and assume that all states are measurable. Most of the software for the overall numerical procedure is involved in the second stage, model identification, where much attention is paid to exciting the system to be measured with the "best" input. That is, it is crucial to produce independent data in the sampling procedure. The final stage, model verification, requires numerous simulation studies and comparative error analyses. One basic course of action in these studies is to emphasize a region of model validity, or region of acceptable performance; the burden of proof, then, lies in showing that as the degree of the model gets larger, so does the region of validity.

It is important to note from the onset that the overshadowing spirit in these modeling studies is to regard the system to be modeled as a "black box" whose dynamics are known to us only by way of state trajectory response curves resulting from various input excitations. We therefore view the system (A.1.1) in this light and attempt to construct a model consisting of a system of nonlinear differential equations, each of which being a truncated series expansion with tensor coefficients.

Consider the nonlinear dynamical equation

$$\dot{x} = f(x, u), \tag{A.2.1}$$

which is the general expression for (A.1.1), for $f: X \times U \to X$, X the n-dimensional state space and U the m-dimensional input space. Using the tensor parameterization in an expansion for the right side of (A.2.1) such as

 $f(x,u) = \begin{bmatrix} L_{10} & L_{01} & L_{20} & L_{11} & L_{02} & L_{30} & \dots \end{bmatrix} x_s, \qquad (A.2.2)$ where the vector x_s contains symmetric powers of x and u,

$$x_8 = (x, u, x \lor x, x \lor u, u \lor u, x \lor x \lor x,...)^T,$$
 (A.2.3)

the true system (A.2.1) may be integrated, and solutions sampled, to form the matrix equation, arising from (A.2.2), as

$$X = LX_{s}. (A.2.4)$$

Here the matrices X and X_8 each have p columns, p being the number of samples taken, where the elements of X are derivative estimates of the states taken along the state trajectories for some input. A lexicographic ordering algorithm is used for forming the matrix X_8 from (A.2.3). Critical parameters in the formulation of (A.2.4) are model degree, sampling interval, the total experiment time, integration stepsize (since the derivatives are to be estimated, a reasonably small stepsize must be used to ensure accuracy), and most important, the choice of input. This leads us to the identification stage of the procedure.

With regard to the matrix Equation (A.2.4), the problem of identifying the model parameters embodied in the matrix L is a straightforward exercise using linear least squares techniques. Indeed, this is one of the primary organizational advantages of the tensor parameterization. The difficulty arises, however, in the procedure used to choose the input to excite enough of the nonlinear modes of the system. A linear combination of sinusoids forms the general structure of the input,

$$u_{j} = \sum_{j=1}^{r} c_{ij} \sin \omega_{ij} t$$
, $j = 1, 2, ..., m$, (A.2.5)

where the cij are input amplitudes and

$$\omega_{ij} = (2\pi f_j)i,$$
 (A.2.6)

for the frequency f_j in hertz. The design of the inputs (A.2.5) is carried out through use of optimization techniques where the quadratic data matrix M (analogous to the information matrix in the information theoretic approach to optimal input design) is made small in some sense. The cost criterion chosen here for accomplishing this is given by

$$J = tr (M^{-1}),$$
 (A.2.7)

so that the design problem may be simply stated as

$$u = \frac{\min}{U} \{J\}, \qquad (A.2.8)$$

for u an element of some admissible set of input trajectories.

A brief discussion of the software package used in this overall scheme will aid in clarifying these procedures. The routine used for formulating Equation (A.2.4) for the least squares identification of L has the following program inputs:

- initial state conditions
- integration stepsize

- model approximation degree
- number of samples (h)
- number of sinusoids per input (r)
- input amplitudes (cij)
- input frequencies (fi)
- optimization option
- optimization method with associated parameters.

The optimization option is a selection parameter by which the user may choose to optimize over input amplitudes, input frequencies, initial state conditions, or any combination thereof. With the optimization method parameter the user has a choice of two techniques each for the constrained or unconstrained optimization problem. For reasons to be explained later, we concern ourselves here with the unconstrained problem such as that given in (A.2.8). One method choice employs the conjugate gradient approach to nonlinear optimization. main advantage of this technique is that it provides a fast rate of convergence without need for storage of any large matrices. But since this indirect optimization approach requires either a good estimate of the cost function gradient or a closed form solution thereof, it is not employed in the example of this case study. Instead, a direct method approach to solution of (A.2.8) is used via a routine from the IMSL support software on the IBM 370/3033. this technique, the quasi-Newton method, it is assumed that the gradient vector and the Hessian matrix exist, but the user is not required to supply formulas for their evaluation. Several free parameters may be specified for use of this routine, such as maximum number of cost function evaluations (to limit CPU time), convergence criterion, and choice of the initial Hessian matrix.

Once all program inputs are specified for the equation formulation, the input design begins with an integration of the true system; a fourth order Runge Kutta algorithm is used for this purpose. The data matrix M is formed initially with the start-up values, and then with values returned by the optimization routine for each subsequent evaluation of the cost function.

Typical optimizations require anywhere from 50 to 150 evaluations of the cost function. The IMSL routine has a built-in stopping mechanism in the event that further progress is impossible. For instance, the iterative procedure may be terminated if it is determined that roundoff errors become dominant. Otherwise, the routine either terminates with convergence (according to a prespecified threshold) or when the user-specified maximum number of cost function calculations has been reached. Along with the value of the objective function, an important quantity to observe in this iterative procedure is the condition number of the data matrix. It is clear that as the matrix approaches a singular matrix the reliability of the optimization deteriorates. For purposes of accuracy, double precision is used throughout.

Once the matrix Equation (A.2.4) is loaded, using data resulting from the designed input, linear least squares is performed to produce the parameters which comprise the nonlinear model. A singular value decomposition routine for solution of simultaneous matrix equations, from the SPEAKEZY library on the IBM 370/3033, is used for this purpose and returns the singular values of $X_{\rm S}$ so that its condition number may be calculated. This serves as another measure of the independence of the sampled data. It should be noted that in this identification step, prior to entering the SPEAKEZY routine, the user is given the option to normalize the data in $X_{\rm S}$ in order to enhance the capabilities of the numerical routines which follow. When this is done, the result from the SPEAKEZY routine (that is, the matrix L in (A.2.4)) is then transformed back in a reverse process.

Recall that the condition number of a matrix is simply the ratio of the largest to smallest singular value of the matrix. An EISPAK SVD routine is used in this application.

The verification stage of the overall procedure is carried out by way of two separate routines in two separate computing environments. To take advantage of the speed and efficiency of the IBM 370/3033, the large exhaustive simulation exercises are done there, while the graphics capabilities and immediate turn-around time opportunities afforded by the DEC PDP 11/44 are utilized for interactive viewing of response curves and hardcopy plotting. In either case, comparative studies are done for model performance.

Simulation parameters (that is, initial state conditions, input amplitudes, and input frequencies) are loaded into a sequential data set for the bulk simulation exercises. The user specifies the system configuration as true system versus either one or two models of the system. These systems are then integrated, again using a fourth-order Runge Kutta algorithm, only now calculations are carried out in single precision for added speed.

If two models are compared against the true system, an error analysis is performed in which a ratio of mean square errors over the integration interval is calculated. To clarify this, consider the discrete version of our problem and suppose that $x_j(k)$ represents the jth state of the true system solution, $xl_j(k)$ the jth state of the first model solution and $x2_j(k)$ the jth state of the second model solution, each at the timepoint k. Then for the jth state the error ratio ϵ_j is given by

$$\varepsilon_{j} = \frac{\sum_{k=1}^{q} (x_{j}(k) - x_{j}(k))^{2}}{\sum_{k=1}^{q} (x_{j}(k) - x_{j}(k))^{2}},$$
(A.2.9)

where q is the total number of time steps in the simulation interval. Therefore, in view of (A.2.9), if

$$\epsilon_{\uparrow} < 1$$
 (A.2.10)

is true, the second model has outperformed the first model in the jth state for that particular set of simulation parameters. In this way a region of model validity may be mapped out when a sufficient amount of simulation results has been generated. Typical simulation tables, as we shall see in the sections to follow, consist of initial state conditions and input amplitudes which increase in some organized fashion about the equilibrium point values. Frequencies of the inputs are also varied, typically from 0 to 10 hertz. Another good test of model region employed throughout the validation studies uses randomly generated (within specified limits) simulation parameters.

Since no one error criterion tells the entire story for any given simulation, another measure is used for these comparative studies. The maximum absolute error over both models compared is multiplied by the total number of time steps q; this value is retained as the error normalization factor (enf). Then the normalized errors

$$\varepsilon l_{j} = \frac{\sum_{k=1}^{q} (x_{j}(k) - x l_{j}(k))^{2}}{\text{enf}}, \qquad (A.2.11)$$

$$\varepsilon 2_{j} = \frac{\sum_{k=1}^{q} (x_{j}(k) - x 2_{j}(k))^{2}}{enf}$$
, (A.2.12)

are printed out for each state, for each simulation, in tabular form. These individual errors alert the user to a situation in which, for instance, an error ratio near unity results but when both model errors are unacceptable. The relative size of the enf, along with the individual normalized errors (A.2.11) and (A.2.12), gives added insight to accompany the error ratio analyses. In general, the larger the enf is, the more relative significance the error ratio carries.

The final and most revealing test for model performance lies in the viewing of state trajectories for simulation parameters within the regions of interest. For this purpose the DEC PDP11/44 computing system is used with the Tektronix 4025A video graphics terminal to display plots for various simulation parameter sets. While the actual time involved for simulations is longer on this machine than on the IBM 370/3033, the trade-off in turnaround time for interactive and hardcopy plotting is worthwhile.

While practical limitations prohibit a truly exhaustive search of regions of validity, the combination of results from simulation studies such as these gives a reasonably complete picture of model performance. An illustration of these ideas appears in the section to follow for the comparative study of a second degree approximation model of the system (A.1.1) versus a linear approximation model, as given in (A.1.2).

A.3 SECOND DEGREE MODELING

We return now to the original system (A.1.1) for this case study, and illustrate the application of the model formulation, identification, and verification as described in the preceding section.

With an integration timestep of 0.005 seconds the system is integrated with initial conditions of 0.005 and -0.005 for x_1 and x_2 , respectively, and sampled every 0.04 seconds for 100 points. A second degree truncation approximation will be considered first, so that the nonlinear model has a total of 28 parameters, 14 for each equation. The input for this model identification consists of a sum of two sinusoids whose amplitudes are held constant at 0.025; after a few trials it was determined that these amplitudes give very good results. Optimization is done over the input frequencies, and the

After 62 evaluations of the cost function convergence is achieved with a cost function reduction of approximately 35%. At this point a local optimization is achieved and the final values of the frequencies are 0.742 and 0.902. The parameters which result from the least squares identification using the data generated with these inputs are shown in Figure A.1, separated according to the individual Lij partitions as depicted in (A.2.2). The eigenvalues of L10 are within one percent of those in (A.1.3), and notice that L10 and L01 embody almost exactly the parameters of the linear approximation (A.1.2). We also make note of the fact that the condition number of the matrix X_S for this application is approximately 19, well within the bounds dictated by limits of double precision calculations. In fact, condition numbers of this magnitude are about as small as can be expected, indicating to a degree the independence of the data achieved with the designed input.

For purposes of comparison a linear model as given in the exact expression (A.1.2) is integrated (forming the solution x1(k) in (A.2.9)) alongside the second degree model of Figure A.1 (forming the solution x2(k) in (A.2.9)). We note that if the identification were to be carried out for the linear approximation as with the second degree approximation, the resulting parameters would be practically identical to those of the exact expression.

In the pages to follow we offer numerous simulation results in the form of tables and plots, as discussed in the preceding section. The scheme for displaying these results, in this section and the next, is explained in the following.

$$L_{10} = \begin{bmatrix} -2.0009 & -2.9978 \\ 0.9998 & 0.9998 \end{bmatrix}$$

$$L_{01} = \begin{bmatrix} 0.0007 & 0.9993 \\ -1.0003 & -0.0000 \end{bmatrix}$$

$$L_{20} = \begin{bmatrix} 0.0019 & -6.0810 & 0.0340 \\ 0.0053 & 0.0009 & -0.0004 \end{bmatrix}$$

$$L_{11} = \begin{bmatrix} -4.0686 & -0.0063 & -0.0305 & -0.0249 \\ -0.0190 & -0.0115 & -0.0122 & -0.0117 \end{bmatrix}$$

$$L_{02} = \begin{bmatrix} -0.0339 & -0.0006 & 0.0049 \\ -0.9991 & 0.0066 & 0.0001 \end{bmatrix}$$

Figure A.1 Second Degree Model

A table is given and accompanied by plots corresponding to one or more simulations, identifiable by the associated number given leftmost in the table. Each table offers different combinations of input amplitudes and frequencies, and initial state conditions. And, as mentioned in the last section, several sets of randomly generated simulation parameters are given. An effort is made to display typical as well as unusual behavior exhibited in the error ratio result.

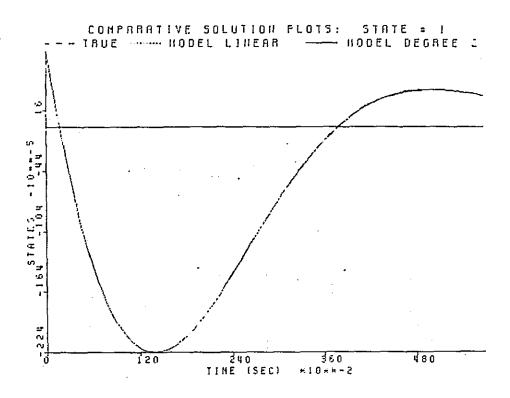
As an example, recalling from (A.2.9) that if the error ratio for a particular state in a given simulation is less than one, we conclude that the second model (in this case the second degree approximation) outperforms the first model (the linear approximation). There may be cases such as simulations 1, 10, 19, and 28 of Tables A.1, A.2, and A.3 for zero input amplitudes in which the error ratios are very large, indicating that the mean square error for the linear model is several orders of magnitude smaller than that of the second degree model. A view of the state trajectories given in Figure A.2 (for simulation l in Table A.1), however, illustrates the fact that all three solution curves--true, linear, and second degree--lie atop one another. make use of another analysis tool we note that the error normalization factor (enf) for each state in this simulation is of the order of 10^{-8} . But the enf for simulation 12 from Table A.1, whose solution curves are shown in Figure A.3, is of the order of 10^{-1} in each state. This serves to reinforce the statement we made earlier concerning the enf: In general, the larger the enf is, the more relative significance the error ratio carries.

As a final note, we point out that the simulations for the first five tables represent data from integrations out to six seconds. For lower frequency inputs though, as in the final three Tables A.6, A.7, and A.8, the simulations are for integrations out to ten seconds.

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1 0.001 0.001 0.000 0.000 0.75 1.00 0.105E+07 0.839E+06 2 0.001 0.001 0.050 0.050 0.75 1.00 0.639E-05 0.658E-04 4 0.001 0.001 -0.050 -0.050 0.75 1.00 0.967E-05 0.951E-04 4 0.001 0.001 -0.050 -0.050 0.75 1.00 0.9667E-05 0.951E-04 6 0.001 0.001 -0.050 0.050 0.75 1.00 0.266E-05 0.951E-04 6 0.001 0.001 0.010 0.150 0.150 0.75 1.00 0.266E-05 0.247E-04 6 0.001 0.001 0.150 0.150 0.75 1.00 0.266E-05 0.247E-02 7 0.001 0.001 0.150 0.150 0.75 1.00 0.390E-03 0.331E-02 8 0.001 0.001 0.150 0.150 0.75 1.00 0.390E-03 0.331E-02 9 0.001 0.001 -0.150 0.150 0.75 1.00 0.613E-03 0.368E-02 10 0.001 -0.001 0.000 0.000 0.75 1.00 0.610E-03 0.388E-02 10 0.001 -0.001 0.000 0.000 0.75 1.00 0.499E+07 0.604E+07 11 0.001 -0.001 0.000 0.000 0.75 1.00 0.499E-05 0.454E-04 12 0.001 -0.001 0.050 0.050 0.75 1.00 0.499E-05 0.454E-04 12 0.001 -0.001 0.050 0.050 0.75 1.00 0.499E-05 0.454E-04 14 0.001 -0.001 0.050 0.050 0.75 1.00 0.492E-05 0.455E-04 14 0.001 -0.001 0.050 0.050 0.75 1.00 0.492E-05 0.455E-04 14 0.001 -0.001 0.050 0.050 0.75 1.00 0.492E-05 0.455E-04 15 0.001 -0.001 0.150 0.150 0.150 0.75 1.00 0.492E-05 0.455E-04 15 0.001 -0.001 0.150 0.150 0.75 1.00 0.575E-03 0.266E-02 16 0.001 -0.001 0.150 0.150 0.75 1.00 0.435E-03 0.333E-02 17 0.001 -0.001 0.150 0.150 0.75 1.00 0.435E-03 0.333E-02 18 0.001 -0.001 0.150 0.150 0.75 1.00 0.435E-03 0.333E-02 18 0.001 -0.001 0.150 0.150 0.75 1.00 0.435E-03 0.333E-02 17 0.001 -0.001 0.150 0.150 0.150 0.75 1.00 0.435E-03 0.333E-02 18 0.001 -0.001 0.150 0.150 0.150 0.75 1.00 0.435E-03 0.333E-02 17 0.001 -0.001 0.150 0.150 0.150 0.75 1.00 0.835E+06 0.666E+02 0.485E-04 0.4	S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
35 -0.001 0.001 -0.150 -0.150 0.75 1.00 0.897F-03 0.478F-02	1 0.001 0.001 2 0.001 0.001 3 0.001 0.001 4 0.001 0.001 5 0.001 0.001 6 0.001 0.001 7 0.001 0.001 9 0.001 0.001 10 0.001 -0.001 11 0.001 -0.001 12 0.001 -0.001 13 0.001 -0.001 14 0.001 -0.001 15 0.001 -0.001 16 0.001 -0.001 17 0.001 -0.001 18 0.001 -0.001 19 -0.001 -0.001 20 -0.001 -0.001 21 -0.001 -0.001 22 -0.001 -0.001 23 -0.001 -0.001 24 -0.001 -0.001 25 -0.001 -0.001 26 -0.001 -0.001 27 -0.001 -0.001 28 -0.001 -0.001 29 -0.001 0.001 31 -0.001 0.001 32 -0.001 0.001 33 -0.001 0.001	0.000 0.000 0.050 -0.050 -0.050 -0.050 -0.050 0.050 -0.150 0.150 0.150 -0.150 -0.150 -0.150 -0.150 -0.050 -0.050 -0.050 -0.050 -0.050 -0.050 -0.150 0.150 -0.150 0.150 -0.150 -0.150 -0.150	0.75 1.00 0.75 1.00	0.105E+07 0.839E+06 0.639E-05 0.658E-04 0.967E-05 0.951E-04 0.536E-05 0.236E-02 0.390E-03 0.311E-02 0.390E-03 0.311E-02 0.499E-05 0.450E-02 0.499E-07 0.454E-04 0.499E-05 0.101E-03 0.499E-05 0.101E-03 0.499E-05 0.101E-04 0.216E-05 0.455E-04 0.216E-03 0.266E-02 0.493E-03 0.333E-02 0.493E-03 0.333E-04 0.545E-03 0.400E-02 0.647E-03 0.430E-04 0.518E-05 0.430E-04 0.518E-05 0.430E-04 0.518E-05 0.430E-04 0.518E-05 0.430E-04 0.546E-05 0.482E-04 0.518E-05 0.482E-04 0.546E-05 0.482E-04 0.546E-05 0.482E-04 0.546E-05 0.482E-04 0.546E-05 0.688E-04 0.546E-05 0.688E-04 0.546E-05 0.688E-04 0.518E-05 0.688E-04 0.518E-05 0.688E-04 0.518E-05 0.688E-04

Table A.1 Linear vs. Degree 2: Constant input frequencies



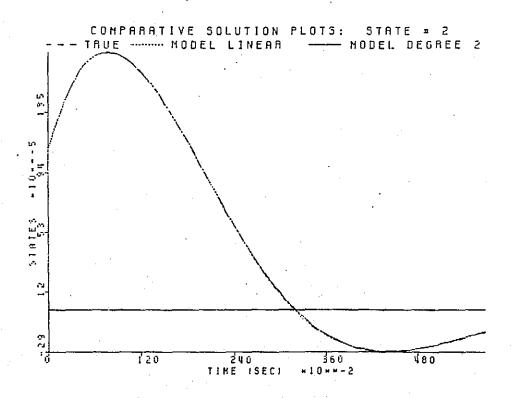
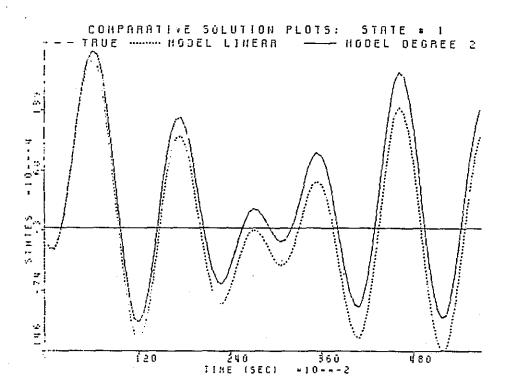


Figure A.2 S#1, Table A.1

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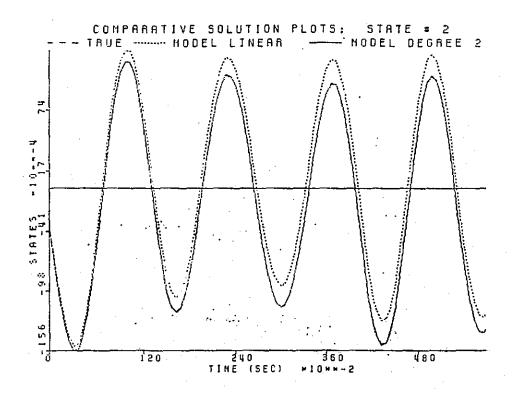
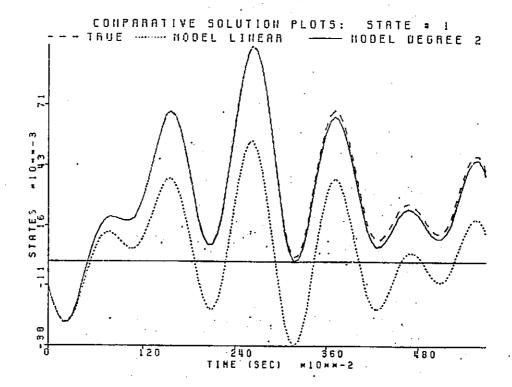


Figure A.3 S#12, Table A.1

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Sŧ	INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
123456789012315678901234567890123456	00000000000000000000000000000000000000	0.050 0.050	00000000000000000000000000000000000000	0.172E+04 0.133E-03 0.133E-04 0.11218E-03 0.133E-04 0.11218E-04 0.11315E-04 0.1238E-04 0.1248E-04 0.1248E-04 0.1248E-04 0.13518E-05 0.14518E-04 0.13518E-05 0.14518E-05 0.14518E-06 0.15136E-07 0.14518E-07 0.15138E-07 0.1513

Table A.2 Linear vs. Degree 2: Constant input frequencies



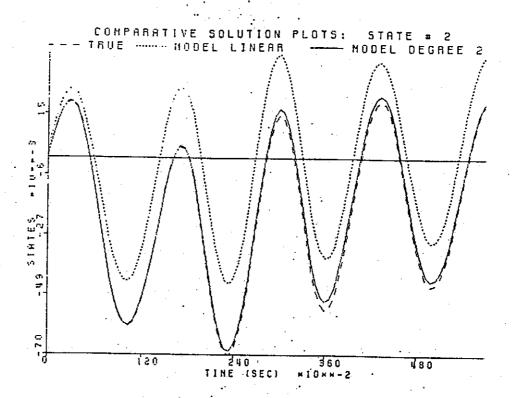
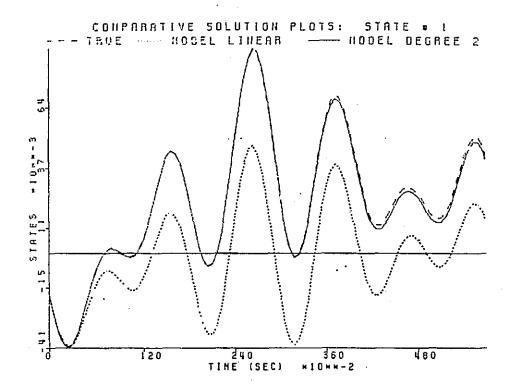


Figure A.4 S#26, Table A.2

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1	0.000 0.050 0.050 -0.050 0.050 -0.050 -0.050 -0.150 0.150 -0.150 0.150 -0.150 -0.150 -0.150	1.00 1.00	142.
35 -0.010 0.010 36 -0.010 0.010	-0.150 -0.150 -0.150 0.150	0.75 1.00 0.75 1.00	0.704E-03 0.389E-02 0.710E-03 0.437E-02

Table 4.3 Linear vs. Degree 2: Constant input frequencies



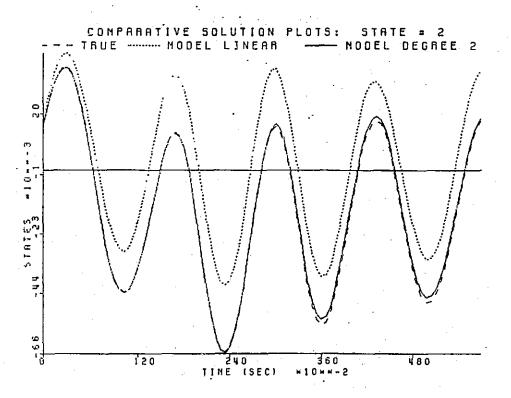
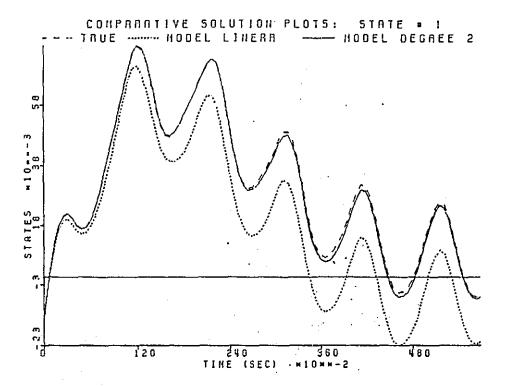


Figure A.5 S#35, Table A.3

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1234555555555555555555555555555555555555	0.100	11.000 11.000	0.466E-03 0.190E-02 0.421E-03 0.206E-02 0.384E-03 0.159E-02 0.374E-03 0.185E-02 0.207E-03 0.926E-03 0.238E-03 0.121E-02 0.276E-03 0.177E-02 0.309E-03 0.873E-03 0.172E-03 0.873E-03 0.172E-03 0.873E-03 0.172E-03 0.873E-03 0.172E-03 0.852E-03 0.172E-03 0.852E-03 0.172E-03 0.852E-03 0.172E-03 0.852E-03 0.172E-03 0.852E-03 0.172E-03 0.852E-02 0.146E-02 0.687E-02 0.146E-02 0.687E-02 0.146E-02 0.551E-02 0.134E-02 0.647E-02 0.134E-02 0.647E-01 0.318E-02 0.135E-01 0.429E-02 0.135E-01 0.429E-02 0.135E-01 0.429E-02 0.198E-01 0.429E-02 0.209E-02 0.259E-02 0.994E-02 0.259E-02 0.994E-01 0.703E-02 0.224E-01 0.703E-02 0.379E-01 0.883E-02 0.224E-01 0.883E-02 0.224E-01 0.979E-02 0.379E-01 0.979E-02 0.379E-01 0.191E-01 0.541E-01 0.193E-01 0.447E-01 0.194E-01 0.553E-01 0.238E-01 0.644E-01 0.195E-01 0.541E-01 0.195E-01 0.541E-01 0.195E-01 0.541E-01 0.195E-01 0.541E-01 0.195E-01 0.643E-01 0.238E-01 0.633E-01 0.278E-01 0.683E-01 0.278E-01 0.713E-01 0.278E-01 0.713E-01 0.278E-01 0.713E-01 0.278E-01 0.713E-01
47 -0.025 -0.025 48 -0.025 -0.025	-0.300 0.300 -0.300 -0.300	2.00 1.00	0.310E-01 0.806E-01

Table A.4 Linear vs. Degree 2: Constant input frequencies



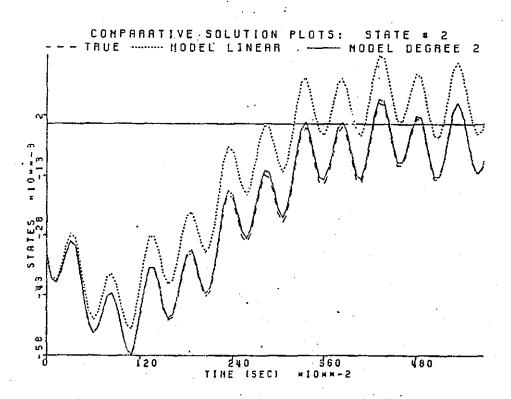


Figure A.6 S#13, Table A.4

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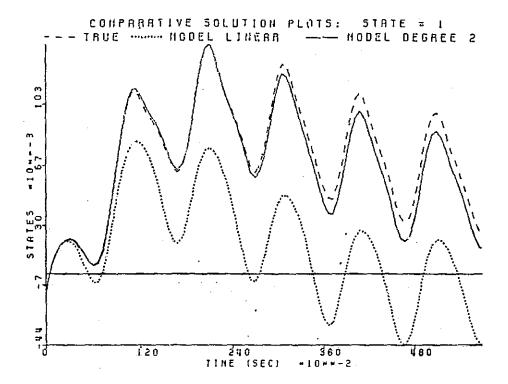
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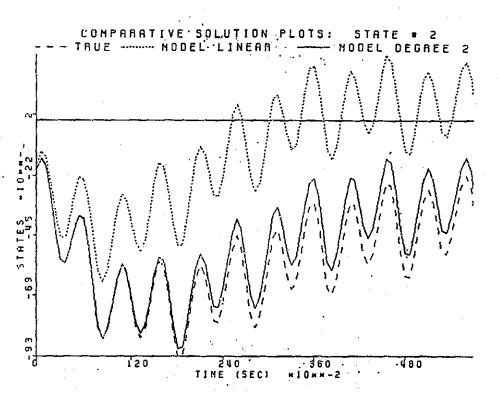
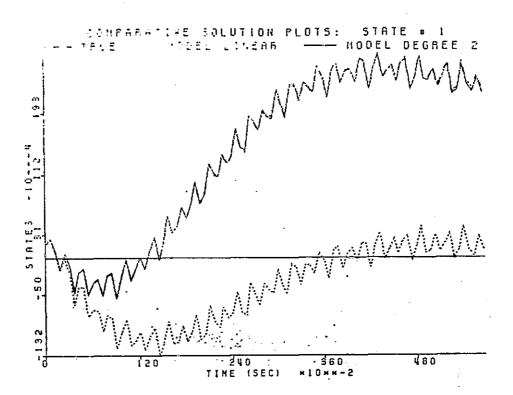


Figure A.7 S#31, Table A.4

S# INITIAL CONDITIONS	S AMPLITUDES	FREQUENCIES	ERROR RATIOS
1 0.005 0.005 5 5 5 5 5 5 5 5 5 5 5 5 5 5	-0.168 -0.1761	37984883448784387445135985214609144567234811554774522 993191354729012144455825511950474457834451480344591 28532554554555045783445545244 29790185650021174648888773322844907869194551109421589128 0083597966521394698888773322844907869194551109421589128 0083597966521154552124948352227092362271561109422443528	0.548E-04 0.200E-03 0.674E-03 0.208E-02 0.226E-04 0.348E-02 0.109E-03 0.243E-03

Table A.5 Linear vs. Degree 2: Constant initial conditions with randomly generated amplitudes and frequencies



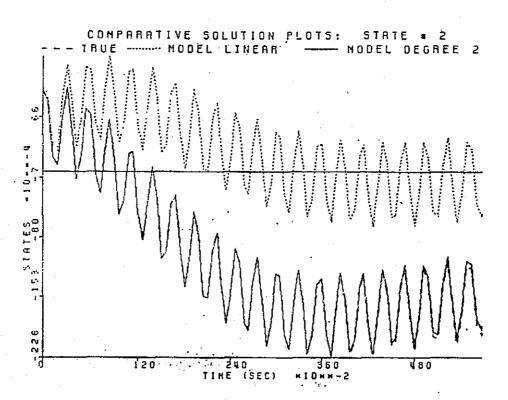
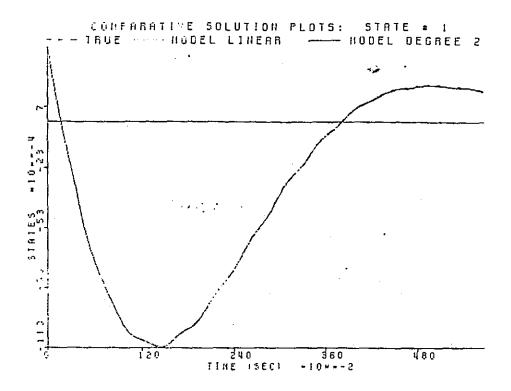


Figure A.8 S#9, Table A.5



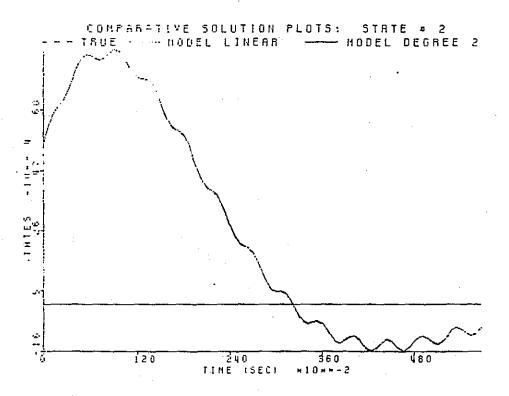
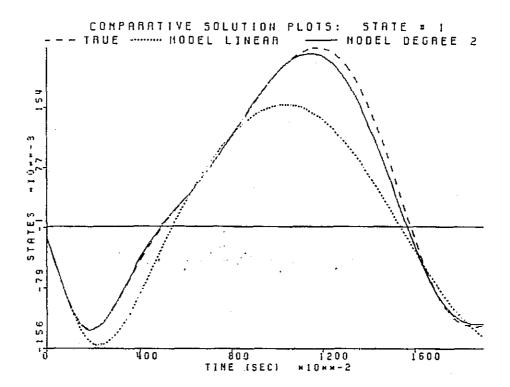


Figure A.9 S#26, Table A.5

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S#	INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
12345678901234567890123456789012	0.001 -0.001 0.001 -0.001 0.010 -0.010 0.010 -0.010	0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.075 -0.050 0.050 -0.050 0.050 -0.050 0.050 -0.075 0.010 -0.075 0.010 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 0.010 -0.075 -0.075 0.010 -0.075 0.010 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075	0.001 0.0025 0.001 0.0025 0.001 0.0025 0.001 0.0025 0.001 0.0025 0.001 0.0025 0.001 0.0025 0.001 0.0025	0.265E-03 0.242E-03 0.251E-03 0.226E-03 0.210E-03 0.193E-04 0.408E-02 0.341E-02 0.372E-02 0.240E-02 0.408E-02 0.240E-02 0.408E-01 0.372E-01 0.408E-01 0.490E-01 0.298E-01 0.490E-01 0.420E-01 0.490E-01 0.149E-03 0.159E-03 0.153E-03 0.236E-03 0.153E-03 0.236E-03 0.169E-03 0.236E-03 0.169E-03 0.159E-03 0.169E-03 0.159E-03 0.169E-03 0.236E-03 0.169E-03 0.236E-03 0.169E-03 0.236E-03 0.169E-03 0.236E-03 0.169E-03 0.236E-03 0.169E-03 0.256E-03 0.256E-03 0.236E-03 0.256E-03 0.236E-03 0.256E-03 0.236E-03 0.256E-03 0.236E-03 0.256E-03 0.256E-03 0.169E-03 0.169E-03 0.175E-03 0.167E-03 0.175E-03 0.125E-03 0.175E-03 0.125E-03

Table A.6 Linear vs. Degree 2: Low frequency inputs



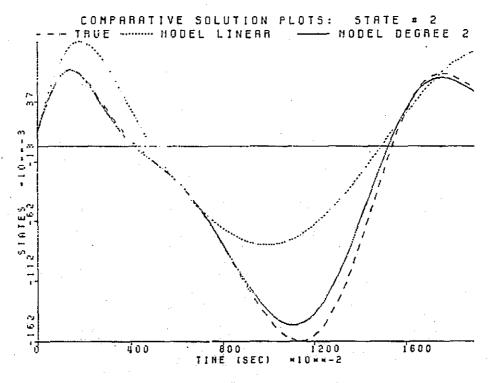


Figure A.10 S#16, Table A.6

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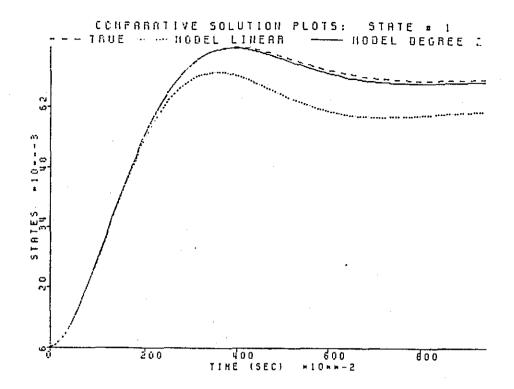
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S# INITIAL CONDITIONS AMPLITUDES FREQUENCIES ERROR RATIOS -0.019 -0.022 -0.019 0.2145E-04 0.125E-05 0.2279E-04 0.155E-06 0.2279E-04 0.155E-06 0.2279E-04 0.155E-06 0.2279E-04 0.155E-06 0.2279E-04 0.155E-06 0.155E-04 0.165E-06 0.161E-02 0.165E-06 0.161E-02 0.1875E-06 0.161E-04 0.1875E-06 0.161E-04 0.1875E-06 0.161E-04 0.1875E-06 0.161E-04 0.1875E-06 0.161E-04 0.1176E-06 0.161E-06 0.1176E-06 0.161E-07 0.1176E-07 0.161E-07 0.1176E-07 0.177E-07 0.1176E-07 0.150E-07 0.1176E-07 0.150E-07 0.1176E-07 0.150E-07 0.1176E-07 0.150E-07 0.1176E-07 0.150E-07 0.1176E-07 0.150E-07 0.1177E-07 0.150E-07 0.1177E-07 0.150E-07 0.1177E-07 0.150E-07 0.1177E-07 0.150E-07 0.1177E-07 0.150E-07 0.1177E-07 0.177E-07 -0.008 0.007 0.00 0.00 0.010 0.024 0.012 0.00 -0.001 0.000 -0.004 -0.008 0.00 -0.004 0.009 -0.018 0.015 0.00 -0.018 0.013 -0.006 0.014 0.012 -0.006 -0.023 0.017 0.023 -0.016 -0.018 -0.008 -0.014 -0.014 -0.013 0.017 -0.006 0.002 0.002 -0.002 -0.001 -0.010 0.00 0.00 0.00 0.00 -0.005 -0.005 0.00 -0.004 -0.000 0.00 0.00 0.00 0.005 -0.001 0.00 0.007 -0.000 0.00 0.00 0.00 0.000 0.022 0,007 -0.006 -0.006 0.022 -0.010 -0.002 0.021 -0.015 -0.020 -0.008 -0.012 0.023 -0.003 0.001 0.015 -0.024 -0.022 -0.015 -0.004 -0.005 0.000 -0.009 0.00 0.006 -0.003 -0.004 0.007 0.002 -0.008 0.002 -0.001 -0.010 0.003 0.003 -0.003 0.00 0.00 18 19 222222222233333 --0.008 0.000 -0.004 -0.006 -0.003 -0.015 0.016 0.000 0.025 0.023 -0.005 0.021 0.001 -0.001 0.003 -0.004 0.005 0.001 0.022 0.005 0.010 0.017 -0.006 -0.013 -0.018 -0.020 0.00 0.00 0.00 -0.001 0.008 -0.002 -0.002 0.00 0.00 0.005 0.006 0.004 -0.001 0.00 0.00 -0.006 -0.001 -0.005 0.002 0.009 0.003 0.00 0.005 0.00 0.00 0.00 0.004 -0.002 0.00 0.00 0.004 0.008 -0.009 -0.003 0.00 0.005 -0.001 0.00 -0.018 -0.020 -0.007 -0.012 0.005 0.008 -0.010 0.009 0.00 0.00 0.005 0.00 9.009 0.00 35 36 37 38 39 0.006 -0.006 0.00 0.00 0.00 ~ 0.009 0.008 -0.003 0.024 0.00 0.008 -0.024 -0.010 0.022 0.009 0.00 0.012 0.019 0.004 0.008 -0.009 0.00 0.00 0.004 -0.003 -0.005 -0.008 -0.002 0.003 0.002 -0.003 0.002 0.010 40 0.012 -0.016 0.00 0.024 -0.004 0.001 0.007 0.00 0.010 -0.009 -0.004 -0.002 -0.002 43 0.00 -0.010 -0.002 -0.023 44 -0.015 0.019 0.00 45 46 47 48 0.004 -0.003 0.00 0.00 0.004 0.006 0.00 0.003 -0.006 0.023 -0.023 -0.015 -0.003 -0.009 0.00 0.00 -0.007 -0.001 0.004 0.00 0.00 0.00 0.009 -0.018 -0.003

Table A.7 Linear vs. Degree 2: Step inputs with randomly generated initial conditions and input amplitudes



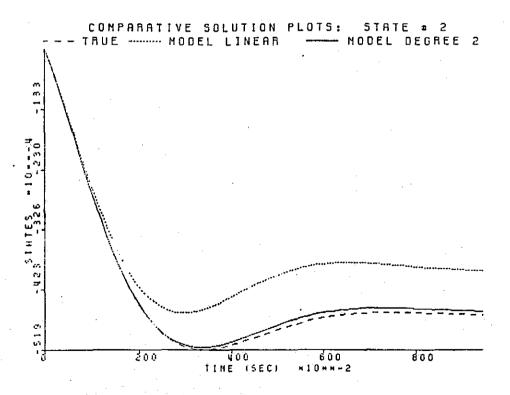
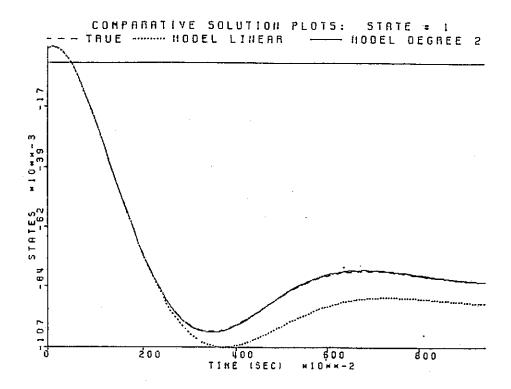


Figure A.11 S#29, Table A.7



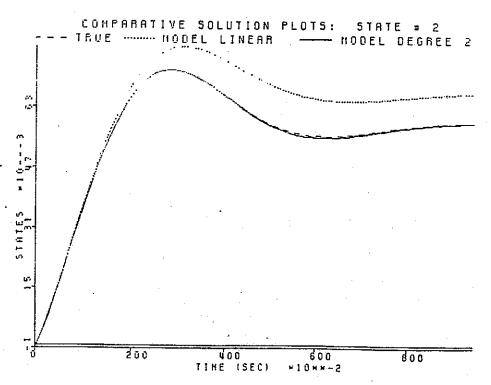


Figure A.12 S#38, Table A.7

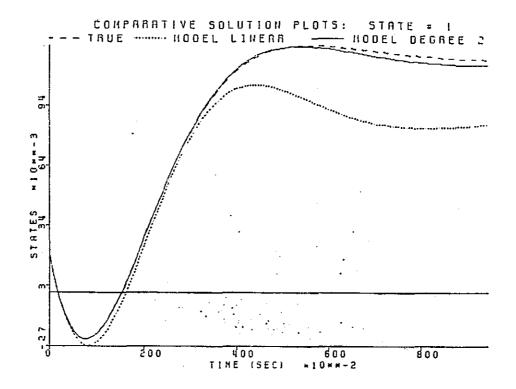
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Sŧ	INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
12345678901234567890123456789012345678901234567890	-0.043 -0.034 -0.030 -0.024 -0.046 -0.014 -0.017 -0.039 -0.047 -0.039 -0.047 -0.034 -0.047 -0.034 -0.047 -0.031 -0.043 -0.001 -0.031 -0.049 -0.031 -0.049 -0.024 -0.039 -0.024 -0.039 -0.025 -0.049 -0.027 -0.039 -0.027 -0.039 -0.029 -0.049 -0.039 -0.049 -0.039 -0.049 -0.039 -0.049 -0.039 -0.049 -0.039 -0.049 -0.039 -0.049 -0.039 -0.049 -0.039 -0.049 -0.049 -0.039 -0.049 -0.039 -0.039 -0.049 -0.039 -0.049 -0.039 -0.049 -0.039 -0.049 -0.039 -0.039 -0.039	-0.070 -0.065 -0.028 -0.028 -0.0172 -0.0104 -0.020 -0.053 -0.018 -0.059 -0.019 -0.059 -0.018 -0.051 -0.019 -0.051 -0.0151 -0.051 -0.015 -0.051 -0.015 -0.051 -0.015 -0.051 -0.015 -0.051 -0.017 -0.051 -0.017 -0.051 -0.018 -0.051 -0.019 -0.034 -0.019 -0.031 -0.019 -0.031 -0.024 -0.031 -0.031 -0.031 -0.031 -0.031 -0.040 -0.031 -0.051 -0.040 -0.051 -0.051 -0.051 -0.055 -0.051 -0.055 -0.051 -0.055 -0.055 -0.055 -0.055 -0.055 -0.055 -0.055 -0.055 -0.055 -0.064 -0.055 -0.064 -0.055 -0.064 -0.064 -0.065 -0.064 -0.065 -0.064 -0.065 -0.065 -0.065 -0.065 -0.		0.483E-03 0.124E-03 0.473E-04 0.275E-04 0.126E-04 0.275E-04 0.126E-04 0.275E-04 0.126E-04 0.275E-04 0.126E-04 0.275E-05 0.126E-04 0.275E-05 0.126E-04 0.275E-05 0.126E-04 0.275E-05 0.126E-04 0.275E-05 0.126E-04 0.275E-05 0.126E-05 0.121E-05 0.126E-07 0.121E-05 0.126E-07 0.1256 -07 0.127E-05 0.1567E-06 0.127E-05 0.1567E-06 0.127E-07 0.1567E-07 0.127E-07 0.1567E-07 0.127E-07 0.1567E-07 0.128E-07 0.17567E-07 0.128E-07 0.17567E-07 0.133E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-07 0.17567E-

Table A.8 Linear vs. Degree 2: Step inputs with randomly generated initial conditions and input amplitudes



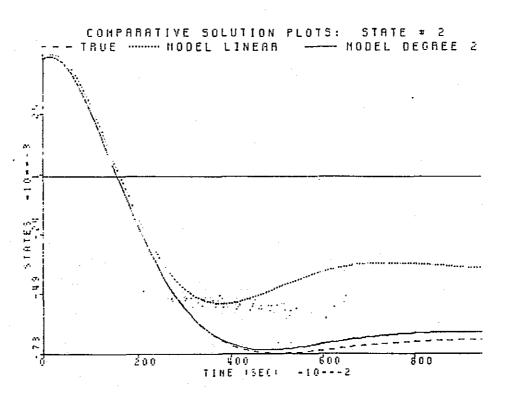


Figure A.13 S#13, Table A.8

A.4 THIRD DEGREE MODELING

The second degree model clearly has a larger region of acceptable performance than the linear approximation, as is evident from the verification studies in the preceding section. In this section we take the identification scheme one step further and construct a third degree truncation approximation using the same procedure.

With an integration timestep of 0.005 seconds the original system is integrated with initial conditions of 0.001 and -0.001 for x1 and x2, respectively, and sampled every 0.04 seconds for 100 points. The input for this model identification consists of a sum of two sinusoids whose amplitudes are held constant at 0.055 and 0.0325 for ul and u2, respectively. Optimization is done over the input frequencies with startup values of 1 and 2 hertz. After 53 evaluations of the cost function convergence is achieved with a cost function reduction of approximately 98%. At this point a local optimum is achieved and the final values of the frequencies are 0.509 and 2.010. For this third degree model with 68 parameters (34 for each equation), the result of the least squares identification using the data generated with these inputs is given in Figure A.14. The condition number of the matrix $X_{\mathbf{S}}$ in this case is about 526, still well within the limits of double precision computations. One would expect that this number be higher than in the second degree model case (recall, there it was 19) since for this identification there are roughly 2.5 times the parameters and thus 2.5 times the data to manipulate.

In this case the comparison studies are done for the model of Figure A.14 versus the second degree model of the previous section. All of the tables to follow, then, have as the first model (that is, solution x1(k) in (A.2.9)) the second degree model of Figure A.1, and have as the second model (that is,

$$L_{10} = \begin{bmatrix} -1.998 & -3.001 \\ 0.999 & 1.000 \end{bmatrix} \quad L_{01} = \begin{bmatrix} 0.002 & 0.998 \\ -1.000 & -0.000 \end{bmatrix}$$

$$L_{20} = \begin{bmatrix} -0.190 & -0.012 & -0.003 \\ 0.036 & 0.002 & 0.000 \end{bmatrix} \quad L_{02} = \begin{bmatrix} -0.004 & 0.007 & -0.012 \\ -0.998 & 0.001 & 0.001 \end{bmatrix}$$

$$L_{11} = \begin{bmatrix} -4.069 & 0.026 & 0.028 & -0.008 \\ 0.021 & 0.003 & 0.003 & 0.004 \end{bmatrix}$$

$$L_{30} = \begin{bmatrix} 0.705 & 1.663 & 3.316 & 0.781 \\ -0.222 & -0.668 & -1.262 & -0.339 \end{bmatrix}$$

$$L_{21} = \begin{bmatrix} 0.621 & -0.614 & -0.252 & -0.765 & 0.447 & -0.675 \\ -0.289 & -0.025 & -0.055 & -0.058 & -0.049 & -0.006 \end{bmatrix}$$

$$L_{12} = \begin{bmatrix} -3.549 & -0.102 & 0.414 & 0.165 & 0.037 & 0.076 \\ -0.172 & 0.960 & -0.044 & -0.056 & -0.035 & -0.013 \end{bmatrix}$$

$$L_{03} = \begin{bmatrix} -0.187 & -0.032 & -0.025 & -0.020 \\ -0.517 & -0.016 & -0.019 & -0.001 \end{bmatrix}$$

Figure A.14 Third Degree Model

solution x2(k)) the third degree model of Figure A.14. Thus, an error ratio less than one indicates that we have succeeded in expanding the region of acceptable behavior by adding more terms in the higher degree approximation.

In the first three tables of this section—A.9, A.10, and A.11 (repeated simulation parameters of Tables A.1, A.2, and A.3)—we find that the third degree model outperforms the second degree model almost uniformly. This of course is an added bonus since these simulation parameters are close to the point of expansion. That is, with these relatively small amplitudes we would hope that the third degree model at least match the performance of the second degree model. Notice, though, that as the input amplitudes get larger the comparative performance for the third degree model improves. One simulation is plotted from each table, revealing that the two models are very close in performance.

The next three tables—A.12, A.13, and A.14—offer simulations with fixed frequencies which progressively stretch the region of operation in terms of the input amplitudes and initial state conditions. While it is clear from the plots which follow that these simulation parameter sets push the second degree model beyond an acceptable region, we note that at the upper end (that is, for input amplitudes of 0.5 and 0.6) we have reached the practical limits of validity for the third degree model performance. Table A.15 offers randomly generated input amplitudes, between -0.3 and 0.3, and input frequencies between 2 and 6 hertz. More of the same type of behavior is observed.

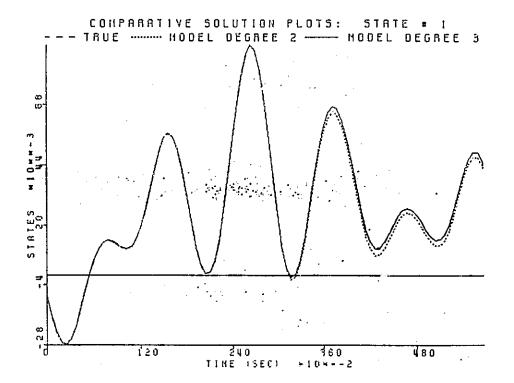
The simulations represented in these first seven tables of this section were all carried out from zero to six seconds, even though some of the plots extend much further. For the final two tables—A.16 and A.17—simulations are carried out to ten seconds. These tables represent simulations for low fre-

quency inputs (including the DC case) and contain identical parameter sets as those in Tables A.6 and A.7 of the last section. These simulations point out the fact that in general, for low frequencies, the third degree model performance is about equal with that of the second degree model. We include plots of two of the worst case results (in terms of the error ratios) in Figure A.30 and Figure A.32. Notice that while the second degree model outperforms the third degree model in these plots, each model exhibits fairly good performance for the worst case. This fact is brought out in the individual error normalization factors as well. In the next section we address this minor difficulty which arises for low frequency inputs.

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S# INITIAL CON	OITIONS AMPLI	TUDES FREC	QUENCIES	ERROR F	RATIOS
1 0.001 (0.001) (0.001	0.001	0.000 0.050	1.00 1.00	618 465 -522 -533 -01 -113E-01 -113E-01 -221E-01 -423 -423 -423 -423 -423 -423 -423 -423 -423 -423 -423 -423 -423 -423 -624 -423 -624 -423 -629 -62	0.22 0.2185 0.2185 0.2185 0.2185 0.2185 0.2185 0.2185 0.2185 0.2185 0.2185 0.2185 0.2185 0.2185 0.2185 0.228 0.228 0.228 0.228 0.228 0.228 0.228 0.228 0.228 0.228 0.228 0.228 0.228 0.288 0.8
35 -0.001	0.001 0.150 0.001 -0.150 0.001 -0.150	-0.150 0. -0.150 0. 0.150 0.	75 1.00 0	0.114E-01 (0.787E-02 (0.218E-01 (0.181E-02 0.200E-02 0.666E-02

Table A.9 Degree 2 vs. Degree 3: Constant input frequencies



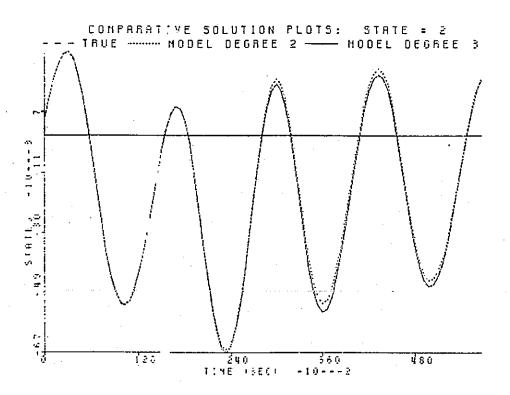
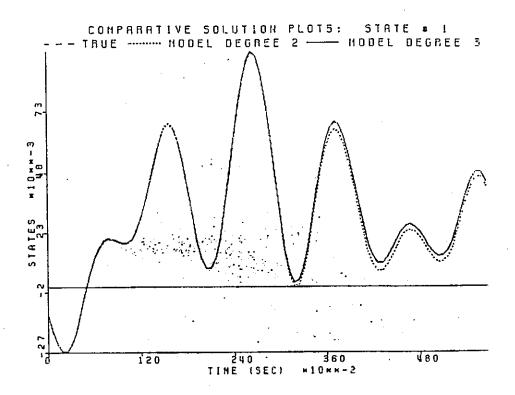


Figure A.15 S#26, Table A.9

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2 0.005 0.005 0.050 0.050 0.75 1.00 0 3 0.005 0.005 0.050 -0.050 0.75 1.00 0	0.923 0.898 0.504 0.788 0.961	0.826 0.602 0.730 0.596
4 0.005 0.005 -0.050 0.75 1.00 0.05 5 0.005 0.005 -0.050 0.75 1.00 0.05 6 0.005 0.005 0.150 0.150 0.75 1.00 0.05 7 0.005 0.005 -0.150 0.150 0.75 1.00 0.05 8 0.005 0.005 -0.150 0.150 0.75 1.00 0.05 9 0.005 0.005 -0.150 0.150 0.75 1.00 0.05 10 0.005 -0.005 0.050 0.050 0.75 1.00 0.05 11 0.005 -0.005 0.050 0.050 0.75 1.00 0.05 12 0.005 -0.005 -0.050 0.050 0.75 1.00 0.05 14 0.005 -0.005 -0.050 0.050 0.75 1.00 0.05 15 0.005 -0.005 -0.150 0.75 1.00 0.05 16 0.005 -0.005 -0.150 <).133E-01).111E-01	1.33 0.670E-02 0.175E-02 0.405E-02 0.846E-02 0.414 0.129 0.634E-01 0.534E-01

Table A.10 Degree 2 vs. Degree 3: Constant input frequencies



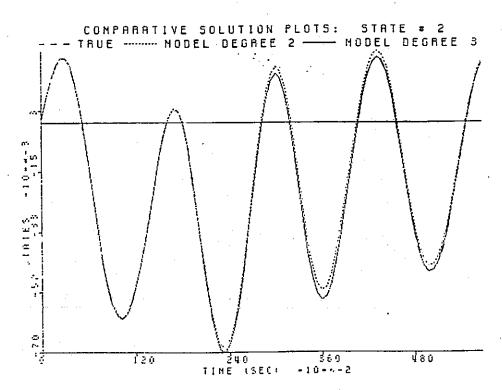
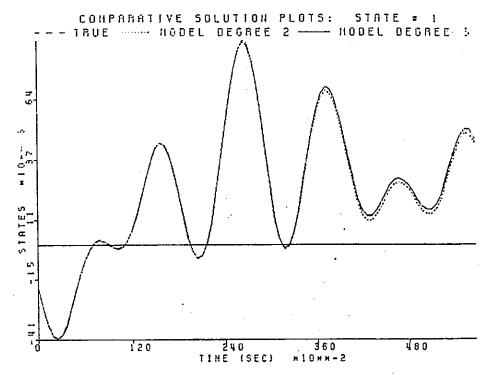


Figure A.16 S#26, Table A.10: Compare to Figure A.4

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Table A. 11 Degree 2 vs. Degree 3: Constant input frequencies

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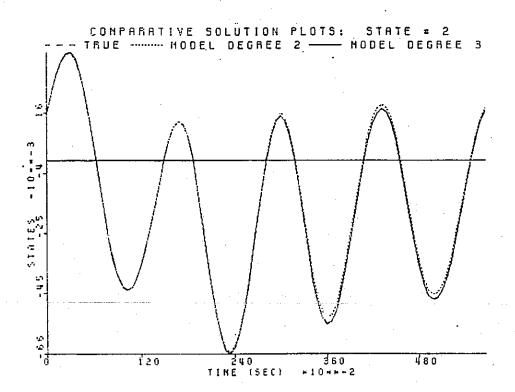
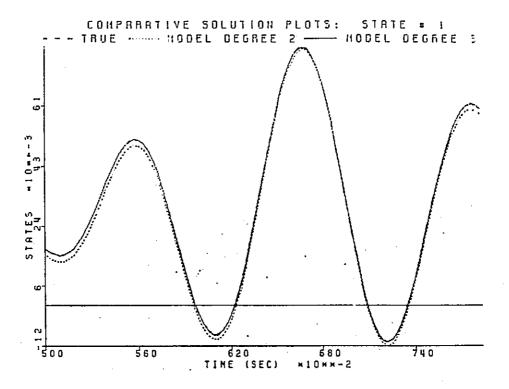


Figure A.17 S#35, Table A.11



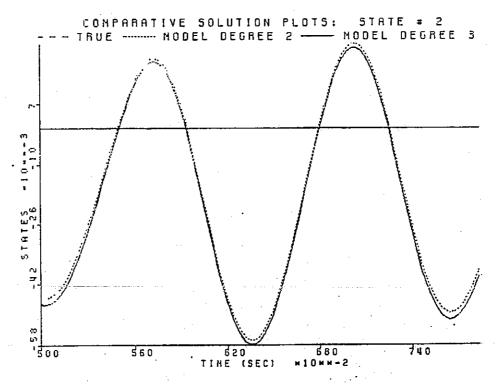


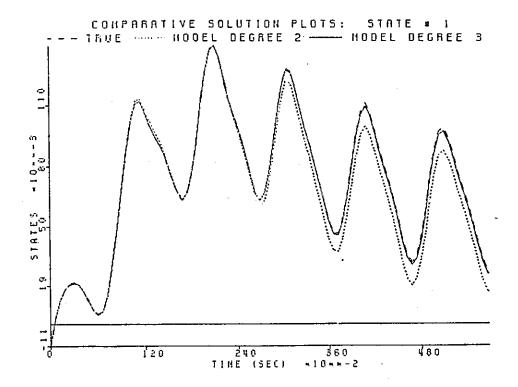
Figure A.18 S#35, Table A.11: Expanded view of Figure A.17

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S# INITIAL CONDIT	IONS AMPLITUDES	FREQUENCIES	ERROR RATIOS
1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0.1000 0.0000 0.0000	1.000000000000000000000000000000000000	0.416 0.402 0.375 0.4285 0.358 0.3285E-02 0.285E-02 0.285E-02 0.285E-02 0.285E-02 0.285E-02 0.285E-02 0.285E-02 0.168E-02 0.1758 0.122 0.1689E-02 0.1689E-02 0.1689E-02 0.203 0.191 0.245E-02 0.2559E-02 0.2756E-02 0.2756E-02 0.2756E-02 0.2756E-02 0.2756E-02 0.2775-02 0.212E-01 0.487E-01 0.5569E-02 0.212E-01 0.429E-01 0.5562E-02 0.213E-01 0.121E-01 0.121FE-01 0.121FE-01 0.1217E-01

Table A.12 Degree 2 vs. Degree 3: Constant input frequencies



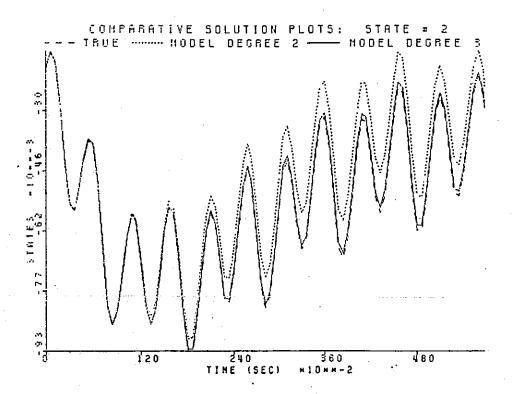
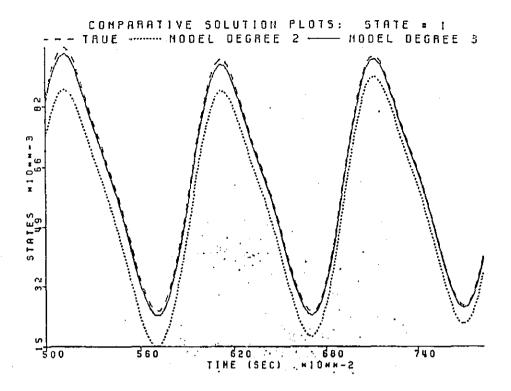


Figure A.19 S#31, Table A.12



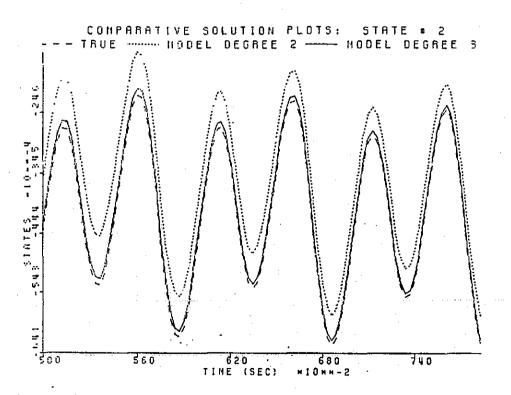
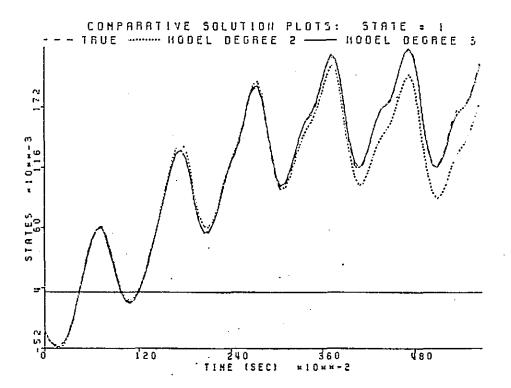


Figure A.20 S#31, Table A.12: Expanded view of Figure A.19



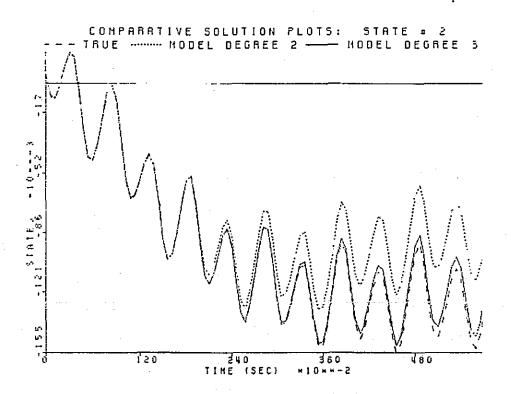
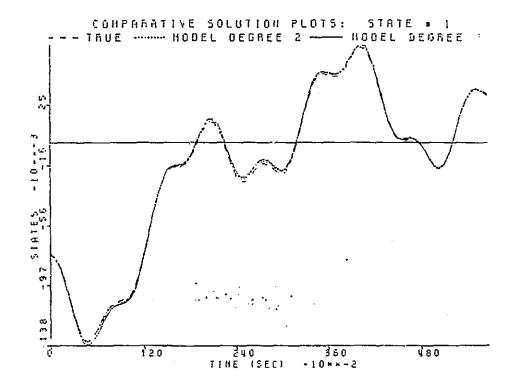


Figure A.21 S#42, Table A.12

Sŧ	INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
123456789012345678901234567890123456789012345678	0.0755555555555555555555555555555555555	0.100	00000000000000000000000000000000000000	3.633 3.224 E - 01 0.3393E - 01 0.425E - 02 0.559E - 02 0.7798E - 01 0.755E - 02 0.798E - 01 0.6552 1.114 1.355 1.114 1.355 1.114 1.355 1.114 1.355 1.114 1.355 1.114 1.355 1.114 1.355 1.114 1.357 1.358 1.350 0.1652E - 01 0.1653E - 01 0.1653E - 01 0.1653E - 01 0.1778E - 02 0.1654E - 01 0.1778E - 02 0.1654E - 01 0.1778E - 01 0.1778E - 01 0.1778E - 02 0.1654E - 01 0.1778E - 01 0.177

Table A.13 Degree 2 vs. Degree 3: Constant input frequencies



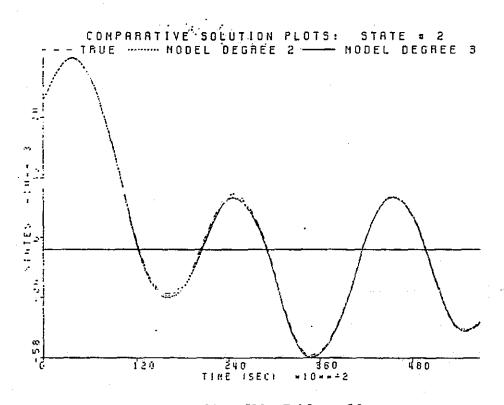
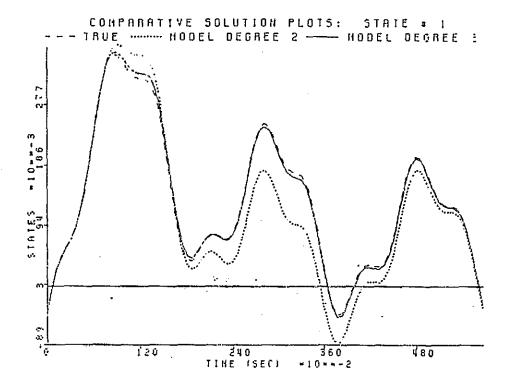


Figure A.22 S#11, Table A.13

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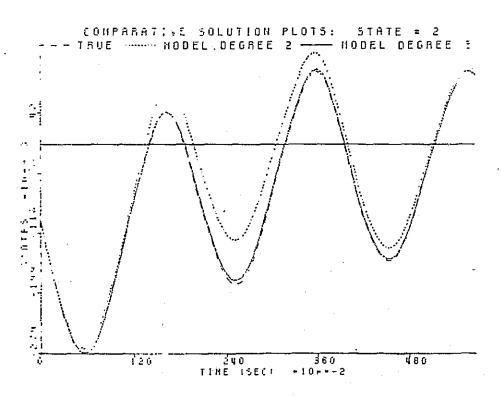


Figure A.23 S#45, Table A.13

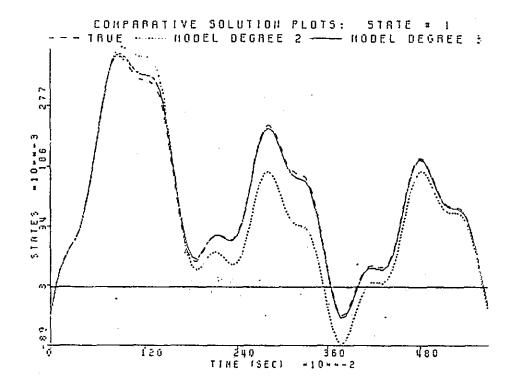
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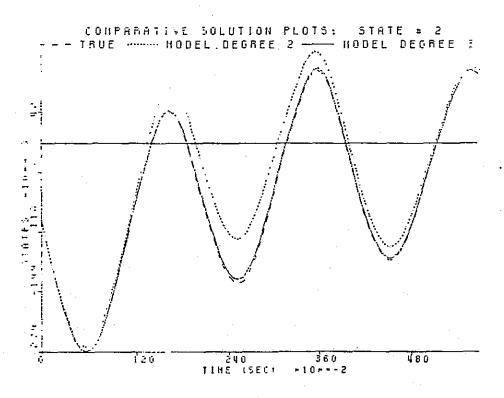
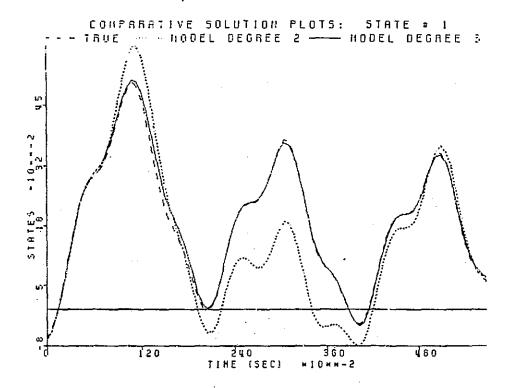


Figure A.23 S#45, Table A.13

S¥	INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
* 123456789012345678901234567890123456789012345	0.075 0.075	0.400 -0.5000 -0.50	\$\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\	0.371E-01 0.118E-01 0.413E-01 0.985E-02 0.530E-01 0.112E-01 0.485E-02 0.145E-02 0.181E-01 0.533E-02 0.180E-01 0.518E-02 0.190E-01 0.518E-02 0.194E-01 0.518E-02 0.194E-01 0.594E-02 0.192E-01 0.594E-02 0.194E-01 0.594E-02 0.192E-01 0.594E-02 0.192E-01 0.594E-02 0.192E-01 0.594E-02 0.194E-01 0.594E-02 0.194E-01 0.594E-02 0.194E-01 0.594E-01 0.895E-01 0.508E-01 0.895E-01 0.321E-01 0.895E-01 0.321E-01 0.895E-01 0.321E-01 0.895E-01 0.321E-01 0.895E-01 0.321E-01 0.895E-01 0.345E-01 0.895E-01 0.301E-01 0.174E-01 0.490E-02 0.316E-01 0.775E-02 0.174E-01 0.150E-01 0.299E-01 0.301E-01 0.299E-01 0.316E-01 0.798E-01 0.316E-01 0.798E-01 0.316E-01 0.798E-01 0.316E-01 0.798E-01 0.788E-01 0.798E-01 0.788E-01
46	-0.075 -0.075	0.600 -0.600	0.50 1.50	0.483E-01 0.346E-01

Table A.14 Degree 2 vs. Degree 3: Constant input frequencies

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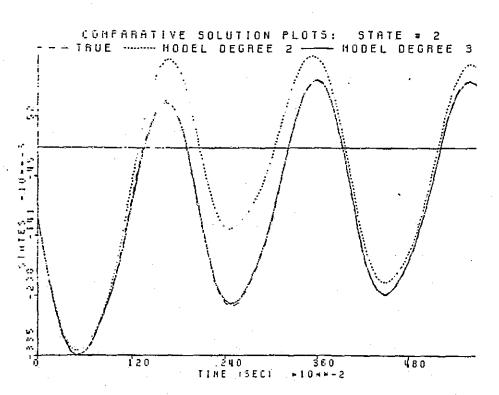
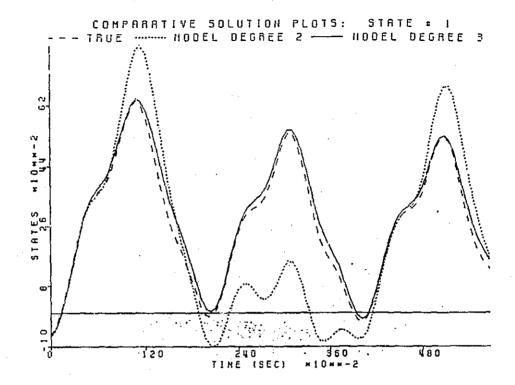


Figure A.24 S#14, Table A.14



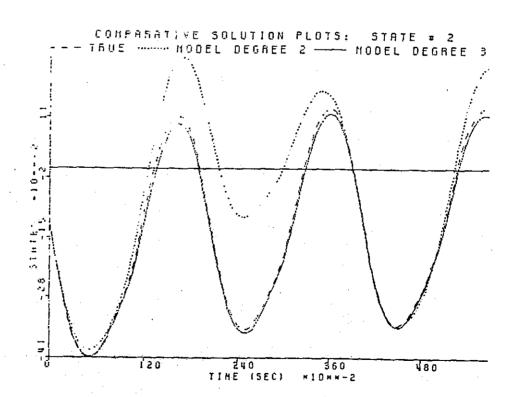
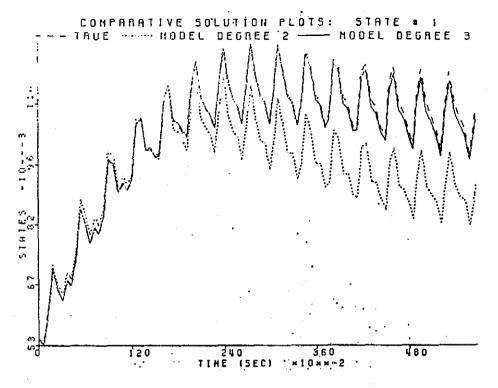


Figure A.25 S#30, Table A.14

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2 0.050 -0.050 -0.087 0.063 4.04 5.67 0.110E-02 0. 3 0.050 -0.050 0.204 -0.277 2.26 3.75 0.124E-01 0. 4 0.050 -0.050 -0.133 -0.138 2.38 3.43 0.456E-02 0. 5 0.050 -0.050 -0.076 0.089 3.43 5.48 0.196E-02 0. 6 0.050 -0.050 0.225 0.215 5.07 2.16 0.103E-01 0.	1.02 0.521E-03
10	00000000000000000000000000000000000000

Table A.15 Degree 2 vs. Degree 3: Constant initial conditions with randomly generated amplitudes and frequencies



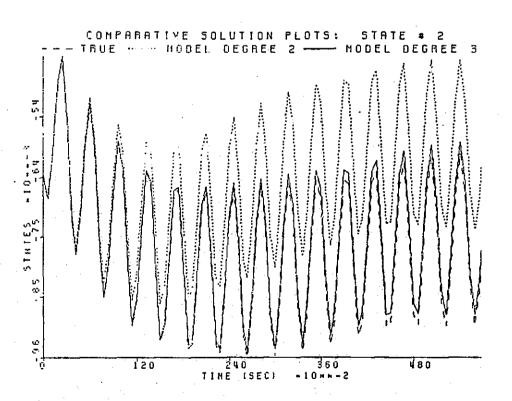
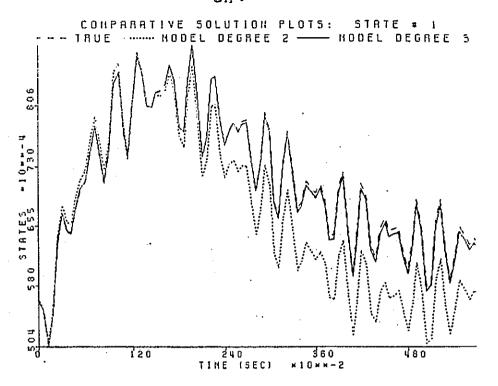


Figure A.26 S#12, Table A.15

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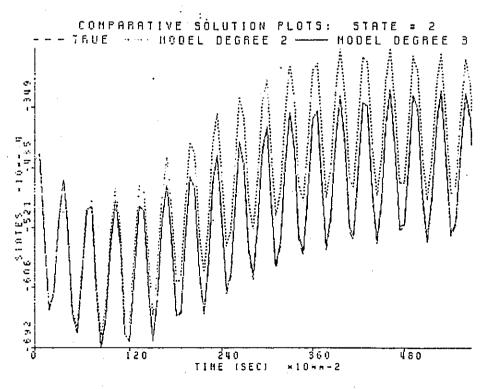
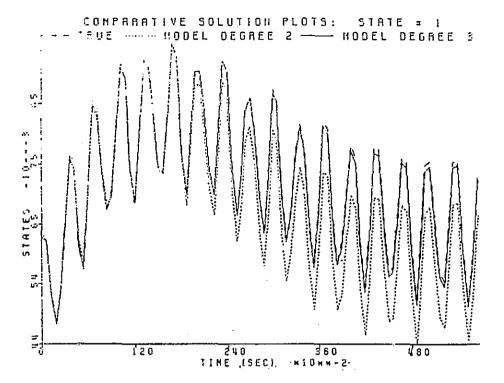


Figure A.27 S#19, Table A.15



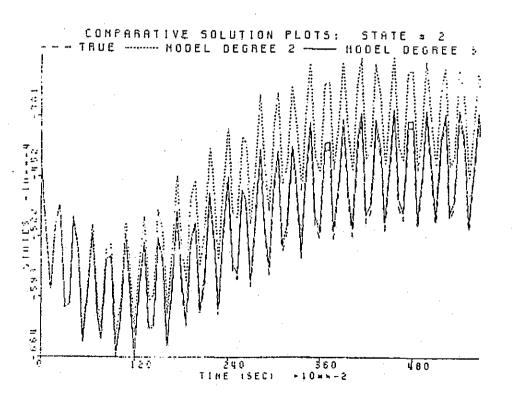
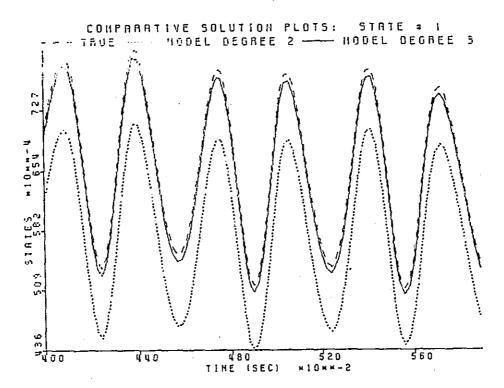


Figure A.28 S#41, Table A.15

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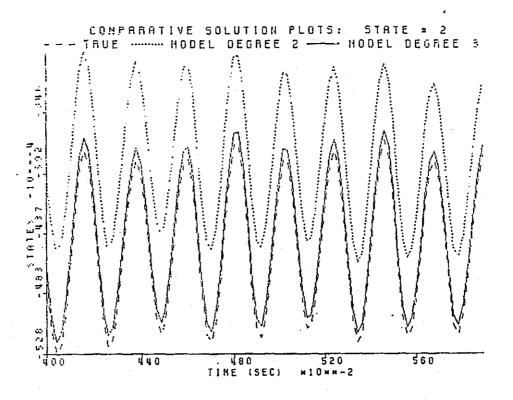
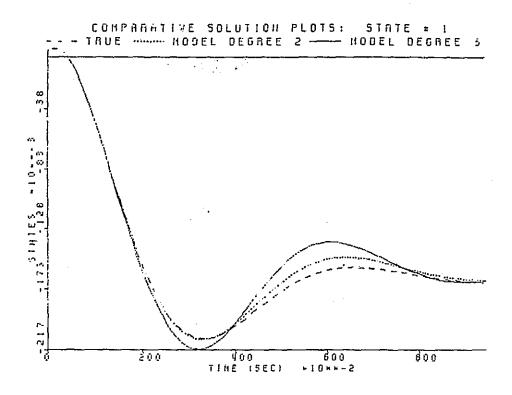


Figure A.29 S#41, Table A.15: Expanded view of Figure A.28

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1 0.001 -0.001 2 0.001 -0.001 3 0.001 -0.001 4 0.001 -0.001 5 0.001 -0.001 6 0.001 -0.001 7 0.001 -0.001 9 0.001 -0.001 10 0.001 -0.001 11 0.001 -0.001 12 0.001 -0.001 13 0.001 -0.001 14 0.001 -0.001 15 0.001 -0.001 16 0.001 -0.001 17 0.010 -0.010 18 0.010 -0.010 19 0.010 -0.010 20 0.010 -0.010 21 0.010 -0.010 22 0.010 -0.010 23 0.010 -0.010 24 0.010 -0.010 25 0.010 -0.010 26 0.010 -0.010 27 0.010 -0.010 28 0.010 -0.010 29 0.010 -0.010	0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.050 -0.050 0.050 -0.050 0.050 -0.075 0.010 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075 -0.075	0.001 0.001 0.001 0.002 0.002 0.003 0.	0.1588777430 30233677430 3023377457447715558333843215779 3023367746133746746137467467467467467467467467467467467467467
31 0.010 -0.010 32 0.010 -0.010	-0.075 -0.075	0.05 0.05	4.37 1.71

Table A.16 Degree 2 vs. Degree 3: Low frequency inputs



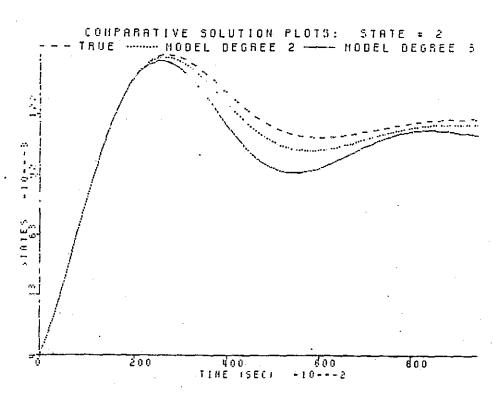
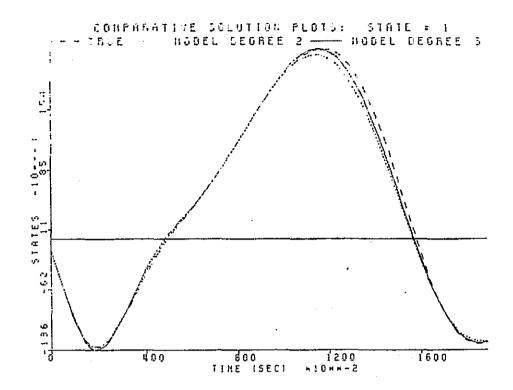


Figure A.30 S#5, Table A.16



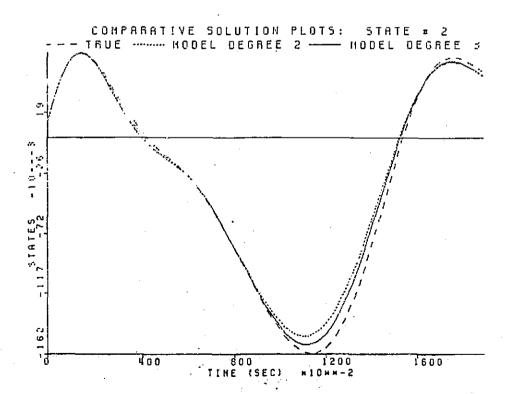


Figure A.31 S#16, Table A.16

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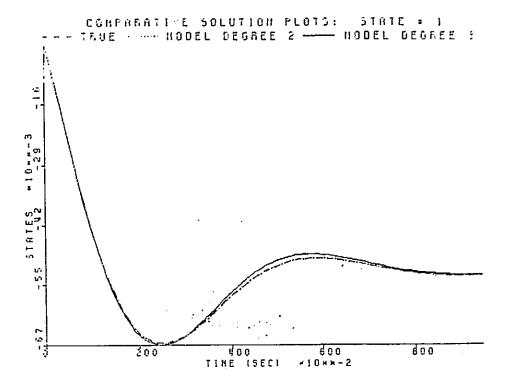
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Sŧ	INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
12345678901234567890123456789012345678901234567890	0.000 0.004 0.0004 0.0007 0.0004 0.0005 0.0004 0.0005 0.00	-0.010 -0.013 -0.012 -0	00000000000000000000000000000000000000	40.8 0.462E-01 9.61 16.5 0.557E-01 0.629E-01 26.2 84.1 233. 0.463E-01 0.171 87.7 0.125 0.182 1.25 0.112 0.111 0.836E-02 0.212E-02 23.7 0.406E-01 0.111 0.836E-02 0.212E-02 23.7 0.406 0.773 123.0 0.371 0.773 123.0 0.371 0.773 123.0 0.261 0.434 0.255.0 0.708E-02 0.108E-01 0.453E-01 0.108E-01

Table A.17 Degree 2 vs. Degree 3: Step inputs with randomly generated initial conditions and input amplitudes



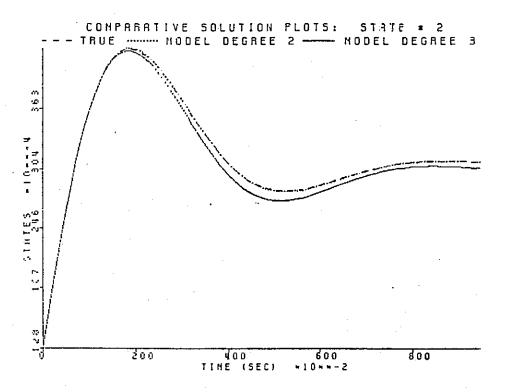


Figure A.32 S#1, Table A.17

A.5 CONCLUSIONS

In this case study we have demonstrated the nonlinear multilinear modeling scheme by way of a two state, two input example system. We began by describing the ways in which the tensor parameterization transforms the nonlinear system into a least squares problem through optimal design of the experimental, measured data. Verification studies were performed with various error analyses to test the region of accentable performance for the resulting model, usually in reference to a lower degree approximation model. While no strict definition of "region" has been given to this point, it is clear that for this example the range of permissible inputs (that is, those which result in acceptable state trajectories) increases as the degree of the approximation model increases. This concept is of course intuitively appealing and in harmony with the implications of Taylor's theorem. The actual implementation is not, however, a trivial exercise when one considers the various software pieces which make up the overall package.

There remain in general several issues concerning the current limitations of the software and numerical routines involved in this procedure. And there remains in particular, for this example, an issue of some concern for model behavior for low frequency inputs. We address this latter issue first by offering a simple means by which the results may be improved.

Recall that the third degree model of the preceding section represents an identification over a four second timespan; 100 samples were taken, spaced evenly in time at 0.04 second intervals. A look at Figure A.30 indicates that the model solution breaks away from the true solution trajectory in each state early in the transient portion of the curve, around the three to four second mark. With this in mind, a re-identification is attempted, this time for

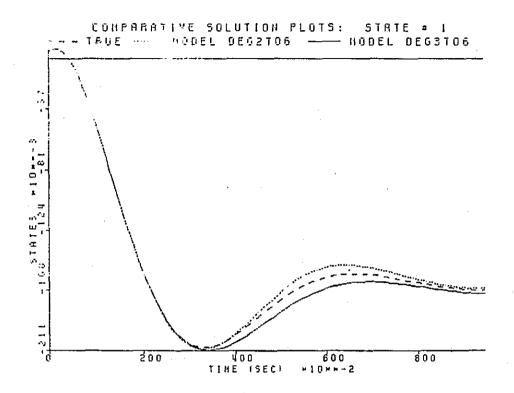
data sampled at 0.06 intervals so that a six second timespan is utilized in an effort to capture more transient information. The final values for the input frequencies from the previous optimization (0.509 and 2.01 hertz) are used now as startup values, everything else being the same. After 50 evaluations of the cost function the final values for the frequencies are 0.351 and 1.33 hertz, the cost function undergoing a 57% reduction. We offer one final table, Table A.18, to illustrate the results for this re-identification, a repeat of Table A.16. For purposes of comparison the ideals cal procedure described above for re-optimization, re-identification is carried out on the second degree model of Section A.3. Both models show modest improvement over their predecessors2, but, as evidenced by Table A.18 compared now against Table A.16, we have improved the low frequency performance of the third degree model a considerable amount. Figure A.33 displays the results of a simulation parameter set which may be compared with Figure A.30. We remind the reader that this simulation represents one of the worst case parameter sets from the table; in the majority of the plots for this table all three curves practically lie atop one another.

As alluded to earlier, the versatility of the software is not currently being exercised. For instance, all of the optimizations for the design of the input are done over the input frequencies, basically because the software package is limited to unconstrained optimizations. Attempting unconstrained optimizations over input amplitudes often results in overly large amplitudes in one or both input channels. Currently, the choice of amplitudes for the identifications is arrived at after a few trial identifications. One method for determining how large these amplitudes should be is to observe the re-

²Resimulation of several tables from Section A.4 illustrate this fact.

S# INITIAL CONDITIONS	AMPLITUDES	FREQUENCIES	ERROR RATIOS
1 0.001 -0.001 2 0.001 -0.001 4 0.001 -0.001 5 0.001 -0.001 6 0.001 -0.001 7 0.001 -0.001 9 0.001 -0.001 10 0.001 -0.001 11 0.001 -0.001 12 0.001 -0.001 13 0.001 -0.001 14 0.001 -0.001 15 0.001 -0.001 16 0.001 -0.001 17 0.001 -0.001 18 0.001 -0.001 18 0.001 -0.010 19 0.010 -0.010 20 0.010 -0.010 21 0.010 -0.010 22 0.010 -0.010 23 0.010 -0.010 25 0.010 -0.010 26 0.010 -0.010 27 0.010 -0.010 28 0.010 -0.010 29 0.010 -0.010 30 0.010 -0.010 31 0.010 -0.010	0.010 -0.075 0.010 -0.075 0.010 -0.075 0.010 -0.050 -0.050 0.050 -0.050 0.050 -0.050 0.050 -0.075 0.010 -0.075	0012501250125012501250125012501250 0000000000	0.757E-02 0.498E-02 0.729E-02 0.560E-02 0.877E-02 0.897E-02 0.177E-01 0.197E-01 1.69 1.77 1.42 1.76 1.03 0.632 0.415 0.650 0.415 0.753 0.536 1.39 0.281 0.379 0.281 0.368 0.287 0.368 0.287 0.368 0.287 0.745E-02 0.569E-02 0.745E-02 0.569E-02 0.7919E-02 0.949E-02 0.192E-01 1.66 1.73 1.41 1.63 0.423 0.633 0.423 1.74 1.03 0.633 0.423 0.5756 0.5427 1.41 0.383 0.651 0.423 0.756 0.5288 0.278 0.228 0.370 0.228 0.298 0.228

Table A.18 A repeat of Table A.16 for improved models: Degree 2 vs. Degree 3



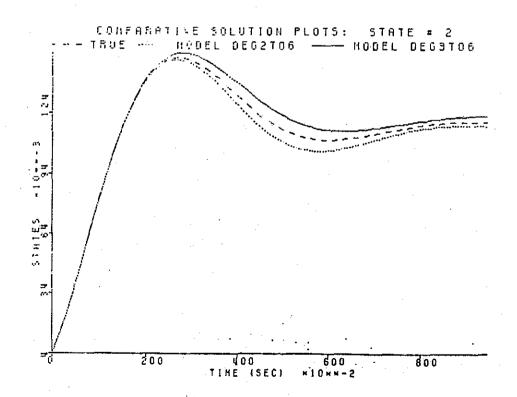


Figure A.33 S#5, Table A.18: Compare to Figure A.30

. Washing

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sulting parameters in the matrices L_{10} and L_{01} ; if these parameters stray too far from the values found for a linear model, local performance will deteriorate.

Another limiting factor of the software currently being used is inherent to the IMSL optimization routine. This quasi-Newton method for finding the optimum parameters in the nonlinear programming problem usually results in local optimum values. Therefore, the choice of startup values for the identification has some influence on the final answer. In the example of this case study, however, the choice of startup frequencies is governed by two factors: 1) knowledge of the speed of response of the system being sampled, and 2) sampling parameters in the identification. We make note of the fact that choice of initial conditions for the identification in this example has a relatively small effect on the result.

We close this case study by commenting on the results of the models discussed in Sections A.3 and A.4 in terms of the so-called region of validity. For simplicity, let us define a "region" in the u₁-u₂ plane as a square whose area is given as the square of some input amplitude. Thus, we speak of regions in terms of maximum input amplitudes.

At low frequencies (that is, up to about 0.1 hertz) both models have roughly the same region of validity, since both perform about the same for input amplitudes out to about 0.075 in magnitude. For higher frequencies, however, we can give some idea for the difference in size of the region in the u1-u2 plane, in terms of amplitudes, in which each model has acceptable behavior. The performance of the second degree model begins to deteriorate for amplitudes out to about 0.2 in magnitude; recall, for example, Figure A.7. The performance of the third degree model is still very good in this region;

compare Figure A.16 to Figure A.4 for amplitudes of magnitude 0.15. In fact, as the plots and simulations from Tables A.14 and A.15 reveal, the third degree model performs well for regions out to amplitude magnitudes of 0.4 and greater. Simply put, then, the third degree model has a region of validity which is at least four times larger than that of the second degree model. A similar measure could be made for the second degree model in terms of the linear approximation. Future software considerations involve a way of graphically displaying these regions as a function of the mean square errors and the error normalization factors. This of course would require a strict definition of when a model's performance is "acceptable".

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